

**Generalized Partial Linear Model with Unknown Link
and Unknown Baseline Functions for Longitudinal Data
(Supplementary Material: conditions and proofs of Theorems)**

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S1 Conditions

To show asymptotic results of our estimators, we need to make following assumptions.

1. Let K_1, K_2, K and \mathcal{K} be bounded and symmetric function with a support $[-1, 1]$ as well as bounded and continuous second derivatives.

We assume that K_1, K_2 and K have orders r_1, r_2 and r_0 , respectively.

$$\text{For example, } \int_{-1}^1 \mu^j K_1(\mu) d\mu = \begin{cases} 1 & j = 0 \\ 0 & 1 \leq j \leq r_1 - 1 \\ \neq 0 & j = r_1 \end{cases} .$$

2. There exists a sequence $\widehat{\beta} = \widehat{\beta}_n$ such that $\|\widehat{\beta} - \beta\| = O_p(n^{-1/2})$.
3. The derivatives $p^{(k_1)}(z), p^{(k_1, k_2)}(z, t), \mu^{(k_1, k_2)}(z, t), \varrho^{(k_1, k_2)}(z, t), \rho^{(k_2)}(t),$

$V^{(k_2)}(t)$, $k_1 = 1, \dots, r_1+1$, $k_2 = 1, \dots, r_2+1$; $q^{(k)}(w)$, $f^{(k)}(w)$, $m^{(k)}(w)$,
 $k = 1, \dots, s+1$ exist and are uniformly bounded over z, t and w .

4. The random vector \mathbf{X} and T are associated with bounded supports.

5. $\frac{\log n}{\sqrt{nh_1^5 h_2}} \rightarrow 0$, $\frac{\log n}{\sqrt{nh_1^3 h_2^3}} \rightarrow 0$ and $\frac{\log n}{\sqrt{nh^3}} \rightarrow 0$.

The assumptions on h_1 , h_2 and h in Theorems 1 and 2 can be satisfied, for example, K_1 is a sixth-order kernel, K_2 and K are fourth-order kernels, $h_1 \propto n^{-1/9}$ and $h_2 \propto n^{-1/5}$, $h \propto n^{-1/5}$. The higher-order kernel for K_1 is needed to assure sufficiently rapid convergence because $\hat{\mu}_1(z, t)$ is a functional of derivatives of K_1 . The high order kernel can be taken from Muller (1984). The initial value of β given in Step 1 is a \sqrt{n} -consistent estimator, hence, Assumption 2 can be satisfied. Assumption 3, the existence of higher-order derivatives of $p(z)$, $p(z, t)$, $\mu(z, t)$, $\varrho(z, t)$, $\rho(t)$, $V(t)$, $q(w)$, $f(w)$ and $m(w)$, is needed to assure that bias terms associated with the kernel estimators $\mu_n(z, t)$, $\mu_{1n}(z, t)$, $\mu_{2n}(z, t)$, $p_n(z, t)$ and $E_n(w)$ vanish sufficiently fast (Horowitz, 1996). Assumption 4 is commonly used in nonparametric literature (Horowitz, 1996; Carroll *et al.*, 1997) and can be relaxed at the expense of more complex proofs.

S2 Proofs

S2.1 Lemmas

Lemmas 1-3 establish asymptotic forms of the convergence of $\widehat{p}(\widehat{z}, t)$, $\widehat{p}y(\widehat{z}, t)$ and their derivatives, which are estimates of $p(z, t)$, $\mu(z, t)p(z, t)$ and their derivatives. These are used in proving Theorem 1 and Lemma 4, which results in a linear approximation to $\frac{\widehat{\mu}_2(\widehat{z}, t)\widehat{p}(\widehat{z}, t)}{\widehat{\mu}_1(\widehat{z}, t)\widehat{p}(\widehat{z}, t)}$ that is used in proving Theorems 1 and 2. Because the arguments used to prove Lemma 3 is essentially same with that in Lemma 2, Lemma 3 is stated without a proof.

Lemma 1: Under the conditions in Supplementary Material S1, we have

$$\begin{aligned} & \frac{(-1)^{k_1+k_2+k_3}}{h_1^{k_1+1}h_2^{k_2+1}} E \left\{ K_1^{(k_1)} \left(\frac{Z_{ij} - z}{h_1} \right) K_2^{(k_2)} \left(\frac{t_{ij} - t}{h_2} \right) \right. \\ & \quad \left. \times S(Z_{ij}, t_{ij}) \right\} = \{S(z, t)p(z, t)\}^{(k_1, k_2)} + O(h_1^{r_1} + h_2^{r_2}), \end{aligned}$$

where $k_1, k_2 = 0, 1, 2$, and S is a function, the derivatives $S^{(r_1+1, r_2+1)}(z, t)$ exist and are uniformly bounded over $\mathbf{x} \in \Omega_1$ and $t \in \Omega_2$. Ω_1 and Ω_2 are the supports of \mathbf{X}_{ij} and t_{ij} , respectively, $z = \mathbf{x}'\boldsymbol{\beta}$.

Proof of Lemma 1: see Horowitz (1996).

Lemma 2: Define

$$\begin{aligned}\widehat{p}(z, t) &= \frac{1}{Nh_1h_2} \sum_{i=1}^n \sum_{j=1}^{n_i} K_1 \left(\frac{\mathbf{X}'_{ij} \widehat{\boldsymbol{\beta}} - z}{h_1} \right) K_2 \left(\frac{t_{ij} - t}{h_2} \right) \\ p_n(z, t) &= \frac{1}{Nh_1h_2} \sum_{i=1}^n \sum_{j=1}^{n_i} K_1 \left(\frac{\mathbf{X}'_{ij} \boldsymbol{\beta} - z}{h_1} \right) K_2 \left(\frac{t_{ij} - t}{h_2} \right), \\ \widehat{py}(z, t) &= \frac{1}{Nh_1h_2} \sum_{i=1}^n \sum_{j=1}^{n_i} Y_{ij} K_1 \left(\frac{\mathbf{X}'_{ij} \widehat{\boldsymbol{\beta}} - z}{h_1} \right) K_2 \left(\frac{t_{ij} - t}{h_2} \right), \\ py_n(z, t) &= \frac{1}{Nh_1h_2} \sum_{i=1}^n \sum_{j=1}^{n_i} Y_{ij} K_1 \left(\frac{\mathbf{X}'_{ij} \boldsymbol{\beta} - z}{h_1} \right) K_2 \left(\frac{t_{ij} - t}{h_2} \right),\end{aligned}$$

Under the conditions in Supplementary Material S1, as $n \rightarrow \infty$, $h_1 \rightarrow 0$ and $h_2 \rightarrow 0$, if $\log n / (nh_1^3 h_2)^{1/2} \rightarrow 0$, we have

$$\begin{aligned}\widehat{p}(\mathbf{x}' \widehat{\boldsymbol{\beta}}, t) &= p_n(\mathbf{x}' \boldsymbol{\beta}, t) + (\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta})' \Gamma^{(100)}(\mathbf{x}' \boldsymbol{\beta}, t, \mathbf{x}) + o_p(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}), \\ \widehat{py}(\mathbf{x}' \widehat{\boldsymbol{\beta}}, t) &= py_n(\mathbf{x}' \boldsymbol{\beta}, t) + (\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta})' \Lambda^{(100)}(\mathbf{x}' \boldsymbol{\beta}, t, \mathbf{x}) + o_p(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}),\end{aligned}$$

uniformly over $\mathbf{x} \in \Omega_1$ and $t \in \Omega_2$, where $\Gamma(z, t, \mathbf{x}) = -p(z, t) [\varrho(z, t) - \mathbf{x}]$, and $\Lambda(z, t, \mathbf{x}) = -p(z, t) [\varrho(z, t) - \mathbf{x}] \mu(z, t)$.

Proof of Lemma 2. We only present the proof of the first equation, and others can be argued in a similar manner. By the mean value theorem of differential calculus

$$\begin{aligned}\widehat{p}(\mathbf{x}' \widehat{\boldsymbol{\beta}}, t) &= \frac{1}{Nh_1h_2} \sum_{i=1}^n \sum_{j=1}^{n_i} K_1 \left(\frac{\mathbf{X}'_{ij} \boldsymbol{\beta} - \mathbf{x}' \boldsymbol{\beta}}{h_1} \right) K_2 \left(\frac{t_{ij} - t}{h_2} \right) \\ &+ \frac{1}{Nh_1^2 h_2} \sum_{i=1}^n \sum_{j=1}^{n_i} K_1^{(1)} \left(\frac{(\mathbf{X}_{ij} - \mathbf{x})' \boldsymbol{\beta}}{h_1} \right) K_2 \left(\frac{t_{ij} - t}{h_2} \right) (\mathbf{X}_{ij} - \mathbf{x})' (\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}) + o_p(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}) \\ &= p_n(\mathbf{x}' \boldsymbol{\beta}, t) + g_n^*(\mathbf{x}, t)' (\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}) + o_p(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}).\end{aligned}\tag{S2.1}$$

By Theorem 2.37 of Pollard (1984, p.34),

$$\sup_{y, \mathbf{X}} |g_n^*(\mathbf{x}, t) - Eg_n^*(\mathbf{x}, t)| = o[(\log n)/(nh_1^3 h_2)^{1/2}]. \quad (\text{S2.2})$$

Using Lemma 1, we get

$$Eg_n^*(\mathbf{x}, t) - \Gamma^{(100)}(z, \mathbf{x}, t) = O(h_1^{r_1} + h_2^{r_2}). \quad (\text{S2.3})$$

By (S2.2), (S2.3) and the conditions on h_1 and h_2 , we have

$$g_n^*(\mathbf{x}, t) - \Gamma^{(100)}(z, t, \mathbf{x}) \rightarrow 0. \quad (\text{S2.4})$$

By (S2.1), (S2.4) and the conditions on h_1, h_2 , Lemma 2 follows.

Lemma 3: Denote $w = \mathbf{x}'\boldsymbol{\beta} + V(t)$, $W_{ij} = \mathbf{X}'_{ij}\boldsymbol{\beta} + V(t_{ij})$, $\widehat{w} = \mathbf{x}'\widehat{\boldsymbol{\beta}} + \widehat{V}(t)$

and $\widehat{W}_{ij} = \mathbf{X}'_{ij}\widehat{\boldsymbol{\beta}} + \widehat{V}(t_{ij})$. Define

$$\begin{aligned} \widehat{f}y(w) &= \frac{1}{Nh} \sum_{i=1}^n \sum_{j=1}^{n_i} Y_{ij} K\left(\frac{\widehat{W}_{ij} - w}{h}\right), \quad fy_n(w) = \frac{1}{Nh} \sum_{i=1}^n \sum_{j=1}^{n_i} Y_{ij} K\left(\frac{W_{ij} - w}{h}\right), \\ \widehat{f}(w) &= \frac{1}{Nh} \sum_{i=1}^n \sum_{j=1}^{n_i} K\left(\frac{\widehat{W}_{ij} - w}{h}\right), \quad f_n(w) = \frac{1}{Nh} \sum_{i=1}^n \sum_{j=1}^{n_i} K\left(\frac{W_{ij} - w}{h}\right). \end{aligned}$$

Under the conditions in Supplementary Material S1, as $n \rightarrow \infty$, $h \rightarrow 0$, if

$\log n/(nh^3)^{1/2} \rightarrow 0$, we have

$$\begin{aligned} \widehat{f}y(\widehat{w}) &= fy_n(w) + (\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta})' \Lambda^{(10)}(w, \mathbf{x}) + \Delta^{(1)}(w) (\widehat{V}(t) - V(t)) \\ &\quad + \frac{1}{Nh^2} \sum_{i=1}^n \sum_{j=1}^{n_i} Y_{ij} K^{(1)}\left(\frac{W_{ij} - w}{h}\right) (\widehat{V}(t_{ij}) - V(t_{ij})) \\ &\quad + o_p(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}) + O_p\left(\frac{\sup_{t \in [0, \tau]} (\widehat{V}(t) - V(t))^2}{h}\right), \end{aligned}$$

$$\begin{aligned} \widehat{f}(\widehat{w}) &= f_n(w) + (\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta})' \Gamma^{(10)}(w, \mathbf{x}) - f^{(1)}(w) \left(\widehat{V}(t) - V(t) \right) \\ &\quad + \frac{1}{Nh^2} \sum_{i=1}^n \sum_{j=1}^{n_i} K^{(1)} \left(\frac{W_{ij} - w}{h} \right) \left(\widehat{V}(t_{ij}) - V(t_{ij}) \right) \\ &\quad + o_p(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}) + O_p \left(\frac{\sup_{t \in [0, \tau]} \left(\widehat{V}(t) - V(t) \right)^2}{h} \right), \end{aligned}$$

uniformly over w , where $\Gamma(w, \mathbf{x}) = -f(w) [q(w) - \mathbf{x}]$, $\Lambda(w, \mathbf{x}) = -f(w) [q(w) - \mathbf{x}] m(w)$, $\Delta(w) = -f(w)m(w)$, $q(w) = E\{\mathbf{X}_{ij} | W_{ij} = w\}$ and $f(w)$ is the density function of W .

Lemma 4: As $n \rightarrow \infty$, if $\log n / (nh_1^3 h_2^3)^{1/2} \rightarrow 0$ and $\log n / (nh_1^5 h_2)^{1/2} \rightarrow 0$, then we have

$$\begin{aligned} \widehat{p}^{(10)}(\mathbf{x}'\widehat{\boldsymbol{\beta}}, t) &= p_n^{(10)}(\mathbf{x}'\boldsymbol{\beta}, t) + (\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta})' \Gamma^{(200)}(\mathbf{x}'\boldsymbol{\beta}, t, \mathbf{x}) + o_p(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}) \\ \widehat{p}^{(01)}(\mathbf{x}'\widehat{\boldsymbol{\beta}}, t) &= p_n^{(01)}(\mathbf{x}'\boldsymbol{\beta}, t) + (\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta})' \Gamma^{(110)}(\mathbf{x}'\boldsymbol{\beta}, t, \mathbf{x}) + o_p(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}), \\ \widehat{py}^{(01)}(\mathbf{x}'\widehat{\boldsymbol{\beta}}, t) &= py_n^{(01)}(\mathbf{x}'\boldsymbol{\beta}, t) + (\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta})' \Lambda^{(110)}(\mathbf{x}'\boldsymbol{\beta}, t, \mathbf{x}) + o_p(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}), \\ \widehat{py}^{(10)}(\mathbf{x}'\widehat{\boldsymbol{\beta}}, t) &= py_n^{(10)}(\mathbf{x}'\boldsymbol{\beta}, t) + (\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta})' \Lambda^{(200)}(\mathbf{x}'\boldsymbol{\beta}, t, \mathbf{x}) + o_p(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}), \end{aligned}$$

uniformly over $\mathbf{x} \in \Omega_1$ and $t \in \Omega_2$, where $\Lambda(z, t, \mathbf{x}) = -p(z, t) [\rho(z, t) - \mathbf{x}] \mu(z, t)$.

Lemma 5: Let

$$\begin{aligned} U(z, t) &= \mu(z, t)p(z, t), \quad U_1(z, t) = \mu^{(10)}(z, t)p(z, t), \quad U_2(z, t) = \mu^{(01)}(z, t)p(z, t), \\ \widehat{U}_1(z, t) &= \frac{\partial \widehat{\mu}(z, t)}{\partial z} \widehat{p}(z, t), \quad \widehat{U}_2(z, t) = \frac{\partial \widehat{\mu}(z, t)}{\partial t} \widehat{p}(z, t), \end{aligned}$$

where $\widehat{\mu}$ and \widehat{p} are obtained by μ_n and p_n with $\boldsymbol{\beta}$ replaced by $\widehat{\boldsymbol{\beta}}$, respectively.

Denote

$$\begin{aligned}\Omega_{n1}(t) &= -\frac{1}{N^2 h_1 h_2^2 \rho(t)} \sum_{i=1}^n \sum_{j=1}^{n_i} \sum_{r=1}^n \sum_{k=1}^{n_r} \{Y_{rk} - \mu(Z_{ij}, t)\} K_1\left(\frac{Z_{rk} - Z_{ij}}{h_1}\right) K_2^{(1)}\left(\frac{t_{rk} - t}{h_2}\right), \\ \Omega_{n2}(t) &= \frac{v(t)}{N^2 h_1^2 h_2 \rho(t)} \sum_{i=1}^n \sum_{j=1}^{n_i} \sum_{r=1}^n \sum_{k=1}^{n_r} \{Y_{rk} - \mu(Z_{ij}, t)\} K_1^{(1)}\left(\frac{Z_{rk} - Z_{ij}}{h_1}\right) K_2\left(\frac{t_{rk} - t}{h_2}\right), \\ \Omega_{n3}(t) &= -\frac{1}{N^2 h_1 h_2 \rho(t)} \sum_{i=1}^n \sum_{j=1}^{n_i} \sum_{r=1}^n \sum_{k=1}^{n_r} \{Y_{rk} - \mu(Z_{ij}, t)\} \left\{ \frac{p^{(01)}(Z_{ij}, t) - v(t)p^{(10)}(Z_{ij}, t)}{p(Z_{ij}, t)} \right\} \\ &\quad \times K_1\left(\frac{Z_{rk} - Z_{ij}}{h_1}\right) K_2\left(\frac{t_{rk} - t}{h_2}\right), \\ \mathcal{Q}(t) &= \frac{1}{\rho(t)} E p(Z_{ij}, t) (\varrho^{(10)}(Z_{ij}, t) \mu_2(Z_{ij}, t) - \varrho^{(01)}(Z_{ij}, t) \mu_1(Z_{ij}, t)),\end{aligned}$$

where $\rho(t)$ is defined in Section 3. Then we have

$$\begin{aligned}\frac{\sum_{i=1}^n \sum_{j=1}^{n_i} \widehat{U}_2(\widehat{Z}_{ij}, t)}{\sum_{i=1}^n \sum_{j=1}^{n_i} \widehat{U}_1(\widehat{Z}_{ij}, t)} &= \frac{\sum_{i=1}^n \sum_{j=1}^{n_i} U_2(Z_{ij}, t)}{\sum_{i=1}^n \sum_{j=1}^{n_i} U_1(Z_{ij}, t)} \\ &= \Omega_{n1}(t) + \Omega_{n2}(t) + \Omega_{n3}(t) + \mathcal{Q}(t)'(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}) + o_p(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}).\end{aligned}$$

Proof of Lemma 5: Since

$$\begin{aligned}\widehat{U}_1(z, t) &= \widehat{p}y^{(10)}(z, t) - \frac{\widehat{p}y(z, t)\widehat{p}^{(10)}(z, t)}{\widehat{p}(z, t)}, \quad U_1(z, t) = U^{(10)}(z, t) - \frac{U(z, t)p^{(10)}(z, t)}{p(z, t)}, \\ \widehat{U}_2(z, t) &= \widehat{p}y^{(01)}(z, t) - \frac{\widehat{p}y(z, t)\widehat{p}^{(01)}(z, t)}{\widehat{p}(z, t)}, \quad U_2(z, t) = U^{(01)}(z, t) - \frac{U(z, t)p^{(01)}(z, t)}{p(z, t)}.\end{aligned}$$

Note that $\mu_2(z, t) = v(t)\mu_1(z, t)$, then we have the expansion,

$$\begin{aligned}
& \frac{\sum_{i=1}^n \sum_{j=1}^{n_i} \widehat{U}_2(\widehat{Z}_{ij}, t)}{\sum_{i=1}^n \sum_{j=1}^{n_i} \widehat{U}_1(\widehat{Z}_{ij}, t)} - \frac{\sum_{i=1}^n \sum_{j=1}^{n_i} U_2(Z_{ij}, t)}{\sum_{i=1}^n \sum_{j=1}^{n_i} U_1(Z_{ij}, t)} \\
= & \frac{1}{N\rho(t)} \sum_{i=1}^n \sum_{j=1}^{n_i} \left\{ \widehat{p}y^{(01)}(\widehat{Z}_{ij}, t) - U^{(01)}(Z_{ij}, t) \right\} \\
& - \frac{v(t)}{N\rho(t)} \sum_{i=1}^n \sum_{j=1}^{n_i} \left\{ \widehat{p}y^{(10)}(\widehat{Z}_{ij}, t) - U^{(10)}(Z_{ij}, t) \right\} \\
& + \frac{1}{N\rho(t)} \sum_{i=1}^n \sum_{j=1}^{n_i} \frac{U(Z_{ij}, t)}{(p(Z_{ij}, t))^2} \left\{ p^{(01)}(Z_{ij}, t) - v(t)p^{(10)}(Z_{ij}, t) \right\} \left(\widehat{p}(\widehat{Z}_{ij}, t) - p(Z_{ij}, t) \right) \\
& - \frac{1}{N\rho(t)} \sum_{i=1}^n \sum_{j=1}^{n_i} \frac{U(Z_{ij}, t)}{p(Z_{ij}, t)} \left(\widehat{p}^{(01)}(\widehat{Z}_{ij}, t) - p^{(01)}(Z_{ij}, t) \right) \\
& + \frac{1}{N\rho(t)} \sum_{i=1}^n \sum_{j=1}^{n_i} \frac{v(t)U(Z_{ij}, t)}{p(Z_{ij}, t)} \left(\widehat{p}^{(10)}(\widehat{Z}_{ij}, t) - p^{(10)}(Z_{ij}, t) \right) \\
& - \frac{1}{N\rho(t)} \sum_{i=1}^n \sum_{j=1}^{n_i} \left\{ \frac{p^{(01)}(Z_{ij}, t) - v(t)p^{(10)}(Z_{ij}, t)}{p(Z_{ij}, t)} \right\} \left(\widehat{p}y(\widehat{Z}_{ij}, t) - U(Z_{ij}, t) \right) + o_p(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}).
\end{aligned}$$

Lemma 5 follows by Lemmas 2-4, the conditions on h_1 and h_2 and some tedious computations.

Lemma 6: Define

$$\begin{aligned}
\Upsilon_n(y) &= \frac{1}{N} \sum_{i=1}^n \sum_{j=1}^{n_i} \int_{y_0}^y \frac{S(u)}{h_p} K_p\left(\frac{\zeta_i - u}{h_p}\right) du \\
\vartheta_n(y) &= \frac{1}{N} \sum_{i=1}^n \sum_{j=1}^{n_i} I(y_0 \leq \zeta_i \leq y) S(\zeta_i)
\end{aligned}$$

where $p = 1, 2$ and the derivative $S^{(r_p+1)}(u)$ exists and uniformly bounded over the bounded set of y . If $nh_p^{2r_p} \rightarrow 0$, then

$$\Upsilon_n(y) - \vartheta_n(y) = o_p(n^{-1/2}),$$

uniformly over the bounded set of y .

Proof of Lemma 6: It is easy to show

$$E[\Upsilon_n(y) - \vartheta_n(y)] = O(h_p^{r_p}). \quad (\text{S2.5})$$

Since $E\left[\int_{y_0}^y \frac{S(u)}{h_p} K_p\left(\frac{\xi_i - u}{h_p}\right) du - I(y_0 \leq \xi_i \leq y)S(\xi_i)\right]^2 = O(h_p)$, by Theorem 2.37 of Pollard (1984), we have

$$\sup_y \|\Upsilon_n(y) - \vartheta_n(y) - E[\Upsilon_n(y) - \vartheta_n(y)]\| = o(h_p^{1/2}(\log n)/n^{1/2}) \quad (\text{S2.6})$$

almost surely. Combining (S2.5) and (S2.6) with the condition on $h_p, p = 1, 2$, we obtain Lemma 6.

S2.2 Proof of Theorem 1

Proof of Theorem 1 relies on two steps. The first step includes an asymptotic expansion of $\widehat{V}(t)$. Combining asymptotic expansions and the estimating equation on β , the asymptotic expansion of $\widehat{\beta}$ is obtained, this is the second step.

Step 1. Note that $v(t) = \mu_2(z, t)/\mu_1(z, t)$, by the result in Lemma 5 and (2.7), we have

$$\begin{aligned} \widehat{V}(t) - V(t) &= \int_0^t \Omega_{n1}(u) du + \int_0^t \Omega_{n2}(u) du \\ &\quad + \int_0^t \Omega_{n3}(u) du + (\widehat{\beta} - \beta)' \pi(t) + o_p(\widehat{\beta} - \beta). \end{aligned} \quad (\text{S2.7})$$

where $\pi(t) = \int_0^t \mathcal{Q}(t)dt$. Define

$$R_n(t) = R_{n1}(t) - R_{n1}(0),$$

where

$$R_{n1}(t) = \frac{1}{Nh_2} \sum_{r=1}^n \sum_{k=1}^{n_r} \frac{\{Y_{rk} - \mu(Z_{rk}, t)\} p(Z_{rk})}{\rho(t)} K_2 \left(\frac{t_{rk} - t}{h_2} \right),$$

and p is the density function of Z . Exchanging the summation in Ω_{n1} and

using the conditions on h_1 , we get

$$\Omega_{n1}(t) = -\frac{1}{Nh_2^2 \rho(t)} \sum_{r=1}^n \sum_{k=1}^{n_r} \{Y_{rk} - \mu(Z_{rk}, t)\} p(Z_{rk}) K_2^{(1)} \left(\frac{t_{rk} - t}{h_2} \right) + O_p(h_1^{r_1}).$$

Hence, by the conditions on h_1 and h_2

$$\begin{aligned} \int_0^t \Omega_{n1}(t) &= \frac{1}{Nh_2} \sum_{r=1}^n \sum_{k=1}^{n_r} \int_0^t \frac{\{Y_{rk} - \mu(Z_{rk}, t)\} p(Z_{rk})}{\rho(t)} dK_2 \left(\frac{t_{rk} - t}{h_2} \right) + O_p(h_1^{r_1}) \\ &= R_n(t) - \frac{1}{Nh_2} \sum_{r=1}^n \sum_{k=1}^{n_r} \int_0^t K_2 \left(\frac{t_{rk} - t}{h_2} \right) \\ &\quad \times \left[\frac{-\mu_2(Z_{rk}, t)}{\rho(t)} - \frac{\{Y_{rk} - \mu(Z_{rk}, t)\} \rho^{(1)}(t)}{\rho^2(t)} \right] p(Z_{rk}) dt + O_p(h_1^{r_1}). \end{aligned}$$

Similarly, we get

$$\begin{aligned} \int_0^t \Omega_{n2}(t) &= \int_0^t \frac{v(t)}{Nh_2 \rho(t)} \sum_{r=1}^n \sum_{k=1}^{n_r} [\{Y_{rk} - \mu(Z_{rk}, t)\} p^{(1)}(Z_{rk}) - \mu_1(Z_{rk}, t) p(Z_{rk})] \\ &\quad \times K_2 \left(\frac{t_{rk} - t}{h_2} \right) dt + O_p(h_1^{r_1}) \\ \int_0^t \Omega_{n3}(t) &= -\int_0^t \frac{1}{Nh_2 \rho(t)} \sum_{r=1}^n \sum_{k=1}^{n_r} \{Y_{rk} - \mu(Z_{rk}, t)\} \left\{ \frac{p^{(01)}(Z_{rk}, t) - v(t) p^{(10)}(Z_{rk}, t)}{p(Z_{rk}, t)} \right\} \\ &\quad \times p(Z_{rk}) K_2 \left(\frac{t_{rk} - t}{h_2} \right) dt + O_p(h_1^{r_1}). \end{aligned}$$

Note that $v(t) = \mu_2(z, t)/\mu_1(z, t)$, hence,

$$\begin{aligned}
& \int_0^t \Omega_{n1}(t) + \int_0^t \Omega_{n2}(t)dt + \int_0^t \Omega_{n3}(t)dt \\
&= R_n(t) + \frac{1}{Nh_2} \sum_{r=1}^n \sum_{k=1}^{n_r} \int_0^t K_2 \left(\frac{t_{rk} - t}{h_2} \right) \frac{1}{\rho(t)} \{Y_{rk} - \mu(Z_{rk}, t)\} \\
&\times \left\{ \frac{\rho^{(1)}(t)p(Z_{rk})}{\rho(t)} + v(t)p^{(1)}(Z_{rk}) - \left\{ \frac{p^{(01)}(Z_{rk}, t) - v(t)p^{(10)}(Z_{rk}, t)}{p(Z_{rk}, t)} \right\} p(Z_{rk}) \right\} dt + O_p(h_1^{r_1}) \\
&= R_n(t) + \frac{1}{N} \sum_{r=1}^n \sum_{k=1}^{n_r} I(t_{rk} \leq t) \frac{\xi(Z_{rk}, t_{rk})}{\rho(t_{rk})} \{Y_{rk} - \mu(Z_{rk}, t_{rk})\} + O_p(h_1^{r_1} + h_2^{r_2}) \quad (\text{S2.8})
\end{aligned}$$

where $\xi(z, t)$ is defined in Section 3.

Step 2. Denote $w = w_0$ and $\hat{w} = \hat{V}(t) + \mathbf{x}'\hat{\boldsymbol{\beta}}$. First, it follows from (2.8) that $\hat{E}(Y|t, \mathbf{x}) \equiv \hat{E}(y|\hat{w}) = \widehat{p}y(\hat{w})/\widehat{p}(\hat{w})$. Applying the results in Lemma 3, we obtain

$$\begin{aligned}
\hat{E}(Y|\hat{w}) - m(w) &= \frac{\widehat{p}y(\hat{w}) - m(w)\widehat{p}(\hat{w})}{\widehat{p}(\hat{w})} \\
&= \frac{\sum_{\ell=1}^n \sum_{s=1}^{n_\ell} (Y_{\ell s} - m(w)) K \left(\frac{V(t_{\ell s}) + \mathbf{X}'_{\ell s}\boldsymbol{\beta} - w}{h} \right)}{Nh f_n(w)} \\
&+ \frac{(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta}_0)' \{ \Lambda^{(10)}(w, \mathbf{x}) - m(w)\Gamma^{(10)}(w, \mathbf{x}) \}}{f(w)} \\
&+ \frac{1}{Nh^2 f(w)} \sum_{\ell=1}^n \sum_{s=1}^{n_\ell} (Y_{\ell s} - m(w)) K^{(1)} \left(\frac{V(t_{\ell s}) + \mathbf{X}'_{\ell s}\boldsymbol{\beta} - w}{h} \right) (\widehat{V}(t_{\ell s}) - V(t_{\ell s})) \\
&- m^{(1)}(w) (\widehat{V}(t) - V(t)) + o_p(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta}) + O_p \left(\frac{\sup_{t \in [0, \tau]} (\widehat{V}(t) - V(t))^2}{h} \right) \quad (\text{S2.9})
\end{aligned}$$

uniformly over \mathbf{x} and t .

Second, noting that $N = O(n)$, the estimate $\hat{\boldsymbol{\beta}}$ is the root of the fol-

lowing equation,

$$\frac{1}{N} \sum_{i=1}^n \sum_{j=1}^{n_i} \{Y_{ij} - \widehat{E}(Y|\widehat{W}_{ij})\} \mathbf{X}_{ij} = 0,$$

which can be rewritten as follows,

$$\frac{1}{N} \sum_{i=1}^n \sum_{j=1}^{n_i} \{Y_{ij} - m(W_{ij})\} \mathbf{X}_{ij} - \frac{1}{N} \sum_{i=1}^n \sum_{j=1}^{n_i} \{\widehat{E}(Y|\widehat{W}_{ij}) - m(W_{ij})\} \mathbf{X}_{ij} = 0. \quad (\text{S2.10})$$

Using (S2.9), exchanging the order of two summations, and applying Condition on h , we obtain

$$\begin{aligned} & \frac{1}{N} \sum_{i=1}^n \sum_{j=1}^{n_i} \{\widehat{E}(Y|\widehat{W}_{ij}) - m(W_{ij})\} \mathbf{X}_{ij} \\ &= \frac{1}{N} \sum_{i=1}^n \sum_{j=1}^{n_i} (Y_{ij} - m(W_{ij})) q(W_{ij}) \\ & - \frac{1}{N} \sum_{i=1}^n \sum_{j=1}^{n_i} \mathbf{X}_{ij} m^{(1)}(W_{ij}) \{q(W_{ij}) - \mathbf{X}_{ij}\}' (\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}_0) \\ & - \frac{1}{N} \sum_{i=1}^n \sum_{j=1}^{n_i} \{(Y_{ij} - m(W_{ij})) q^{(1)}(W_{ij}) - m^{(1)}(W_{ij}) [q(W_{ij}) - \mathbf{X}_{ij}]\} (\widehat{V}(t_{ij}) - V(t_{ij})) \\ & + O_p(h^{r_0}) + o_p(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}) + O_p\left(\frac{\sup_{t \in [0, \tau]} (\widehat{V}(t) - V(t))^2}{h}\right). \end{aligned} \quad (\text{S2.11})$$

Substituting (S2.7) and (S2.8) into (S2.11), we get

$$\begin{aligned}
& \frac{1}{N} \sum_{i=1}^n \sum_{j=1}^{n_i} \{ \widehat{E}(Y | \widehat{W}_{ij}) - m(W_{ij}) \} \mathbf{X}_{ij} \\
&= \frac{1}{N} \sum_{i=1}^n \sum_{j=1}^{n_i} (Y_{ij} - m(W_{ij})) \left\{ q(W_{ij}) + \frac{\xi(Z_{ij}, t_{ij})}{\rho(t_{ij})} + \frac{\kappa(t_{ij})p(Z_{ij})}{\rho(t_{ij})} \right\} \\
&+ \frac{1}{N} \sum_{i=1}^n \sum_{j=1}^{n_i} \{ m^{(1)}(W_{ij}) [q(W_{ij}) - \mathbf{X}_{ij}] \pi'(t_{ij}) - \mathbf{X}_{ij} m^{(1)}(W_{ij}) \{q(W_{ij}) - \mathbf{X}_{ij}\}' \} (\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}) \\
&+ o_p(n^{-1/2} + \|\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}\|) + O_p \left(\frac{\sup_{t \in [0, \tau]} (\widehat{V}(t) - V(t))^2}{h} \right), \tag{S2.12}
\end{aligned}$$

where $\kappa(t) = E \{ \{ m^{(1)}(W) [q(W) - q(W, t)] \} f(t_{k\ell} | W) \}$, and $f(\cdot | w)$ is the conditional density of T given $W = w$. Substituting (S2.12) into (S2.10),

Theorem 1 follows.

S2.3 Proof of Theorem 2

By (S2.7), (S2.8), the conditions on h_1, b and Theorem 1, we have

$$\begin{aligned}
\widehat{V}(t) - V(t) &= R_{n1}(t) + \frac{1}{Nb} \sum_{r=1}^n \sum_{k=1}^{n_r} \int_0^t \mathcal{K} \left(\frac{t_{rk} - t}{b} \right) \frac{1}{\rho(t)} \{ Y_{rk} - \mu(Z_{rk}, t_{rk}) \} \xi(Z_{rk}, t) dt \\
&+ \frac{1}{Nb} \sum_{r=1}^n \sum_{k=1}^{n_r} \int_0^t \mathcal{K} \left(\frac{t_{rk} - t}{b} \right) \frac{1}{\rho(t)} \mu^{(01)}(Z_{rk}, t) (t_{rk} - t) \xi(Z_{rk}, t) dt \\
&+ \frac{1}{Nb} \sum_{r=1}^n \sum_{k=1}^{n_r} \int_0^t \mathcal{K} \left(\frac{t_{rk} - t}{b} \right) \frac{1}{\rho(t)} \mu^{(02)}(Z_{rk}, t) (t_{rk} - t)^2 \xi(Z_{rk}, t) dt / 2 \\
&+ O_p(h_1^{r_1}) + o_p(b^2) \\
&= R_{n1}(t) + Q_n(t) + a_0 b^2 \tau(t) + O_p(h_1^{r_1}) + o_p(b^2), \tag{S2.13}
\end{aligned}$$

where $\tau(t) = \int_0^t \frac{\varphi_1^{(10)}(t,t) + \varphi_2(t)/2}{\rho(t)} dt$, $\varphi_1(t, t)$ and $\varphi_2(t)$ are defined in Section 3,

$$Q_n(t) = \frac{1}{Nb} \sum_{r=1}^n \sum_{k=1}^{n_r} \int_0^t \mathcal{K} \left(\frac{t_{rk} - t}{b} \right) \frac{1}{\rho(t)} \{Y_{rk} - \mu(Z_{rk}, t_{rk})\} \xi(Z_{rk}, t) dt.$$

Since

$$\begin{aligned} \text{Var}\{Q_n(t)\} &= \frac{1}{N} E \left(\text{Var} [Y_{rk} | Z_{rk}, t_{rk}] \left\{ \int_0^t \mathcal{K} \left(\frac{t_{rk} - t}{b} \right) \frac{\xi(Z_{rk}, t)}{b\rho(t)} dt \right. \right. \\ &\quad \left. \left. - I(0 \leq t_{rk} \leq t) \frac{\xi(Z_{rk}, t_{rk})}{\rho(t_{rk})} + I(0 \leq t_{rk} \leq t) \frac{\xi(Z_{rk}, t_{rk})}{\rho(t_{rk})} \right\}^2 \right), \end{aligned}$$

and $E \left\{ \int_0^t \mathcal{K} \left(\frac{t_{rk} - t}{b} \right) \frac{\xi(Z_{rk}, t)}{b\rho(t)} dt - I(0 \leq t_{rk} \leq t) \frac{\xi(Z_{rk}, t_{rk})}{\rho(t_{rk})} \right\}^2 = O(b)$, we have

$$\begin{aligned} \text{Var}\{Q_n(t)\} &= O\left(\frac{b}{N}\right) + \frac{1}{N} E \left(\text{Var} [Y_{rk} | Z_{rk}, t_{rk}] \left\{ I(0 \leq t_{rk} \leq t) \frac{\xi(Z_{rk}, t_{rk})}{\rho(t_{rk})} \right\}^2 \right) \\ &= O\left(\frac{1}{N}\right). \end{aligned}$$

This coupled with $E\{Q_n(t)\} = 0$ results in

$$Q_n(t) = O(n^{-1/2}). \quad (\text{S2.14})$$

Now, we consider $R_{n1}(t)$. It can be shown that

$$ER_{n1}(t) = \frac{b^2}{\rho(t)} E \left\{ \mu^{(01)}(Z, t)p^{(01)}(Z, t) + \mu^{(02)}(Z, t)p(Z, t)/2 \right\} \int x^2 \mathcal{K}(x) dx, \quad (\text{S2.15})$$

and

$$\begin{aligned} \text{Var}\{R_{n1}(t)\} &= \frac{1}{Nb^2} E \left\{ \frac{\{Y_{rk} - \mu(Z_{rk}, t)\} p(Z_{rk})}{\rho(t)} \mathcal{K} \left(\frac{t_{rk} - t}{b} \right) \right\}^2 (1 + o_p(1)) \\ &= \frac{1}{Nb} \frac{E[H(Z, t)p(Z, t)p(Z)]}{\rho^2(t)} \int_x \mathcal{K}^2(x) dx (1 + o_p(1)), \quad (\text{S2.16}) \end{aligned}$$

where $H(z, t) = \text{Var}(Y_{rk} | Z_{rk} = z, t_{rk} = t)$. Hence, Theorem 2 follows from (S2.13), (S2.14), (S2.15) and (S2.16).