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## Mark-Specific Proportional Hazards Model with Multivariate Continuous Marks and Its Application to HIV Vaccine Efficacy Trials

Yanqing Sun

Department of Mathematics and Statistics

University of North Carolina at Charlotte, Charlotte, NC 28223

 $\mathrm{Mei}\ \mathrm{Li}^*$ 

School of Public Health

Zhejiang University, Hangzhou, China

Peter B. Gilbert

Department of Biostatistics, University of Washington

and Fred Hutchinson Cancer Research Center, Seattle, WA 98109

Supplementary Material

This Supplementary Material presents the proofs for Theorem 2.1, Theorem 2.2 and Theorem

3.1 of the paper. Please refer to the main paper for notations not defined here.

Lemma 1 Under conditions (A.1)-(A.3),  $S_k^{(j)}(t, v, \bar{\beta})$  converges to  $s_k^{(j)}(t, v, \bar{\beta})$  in probability uniformly in  $(t, v, \bar{\beta}) \in [0, \tau] \times [0, 1]^2 \times \mathcal{B}$  as  $n \to \infty$ , for j = 0, 1, 2 and  $1 \leq k \leq K$ .

Proof of Lemma 1.

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<sup>\*</sup>To whom correspondence should be addressed.

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Let  $S_{ki}^{(j)}(t, v, \bar{\beta}) = Y_{ki}(t) \exp\{\bar{\beta}^T \tilde{Z}_{ki}(t, v)\}\tilde{Z}_{ki}(t, v)^{\otimes j}$ . Then  $S_k^{(j)}(t, v, \bar{\beta}) = n_k^{-1} \sum_{i=1}^{n_k} S_{ki}^{(j)}(t, v, \bar{\beta})$ . We prove the lemma for the case when j = 0. The proofs for j = 1 and 2 follow similarly. Let  $\omega_{ki} = (X_{ki}, Z_{ki})$ . Then  $\omega_{ki}$ ,  $i = 1, \ldots, n_k$ , is a random sample from a probability distribution  $P_k$  on a measurable space  $(\mathcal{X}_k, \mathcal{A}_k)$ , where  $\mathcal{X}_k = [0, \tau] \times \mathbb{R}^p$  and  $\mathcal{A}_k$  is its Borel  $\sigma$ -field. Let  $\mathcal{F}$  be the class of all coordinate projections  $f_{t,v,\bar{\beta}}(\omega_{ki}) : \mathcal{X}_k \longrightarrow \mathbb{R}$ , where  $f_{t,v,\bar{\beta}}(\omega_{ki}) = S_{ki}^{(0)}(t, v, \bar{\beta})$ , for  $(t, v, \bar{\beta}) \in [0, \tau] \times [0, 1]^2 \times \mathcal{B}$ . Then  $S_k^{(0)}(t, v, \bar{\beta}) = n_k^{-1} \sum_{i=1}^{n_k} f_{t,v,\bar{\beta}}(\omega_{ki})$ . Let  $\|f_{t,v,\bar{\beta}}\|_{P_k,r} = (P_k|f_{t,v,\bar{\beta}}|^r)^{1/r} = (E_k|S_{ki}^{(0)}(t, v, \bar{\beta})|^r)^{1/r}$  be  $L_r(P_k)$ -norm of  $f_{t,v,\bar{\beta}}$ . Next, we show that  $\mathcal{F}$  is Glivenko-Cantelli (van der Vaart, 1998). Since  $Z_{ki}(\cdot)$  is of bounded variation, for simplicity we assume that  $Z_{ki}(\cdot)$  is an nonnegative monotone increasing processes plus a constant. In this case, the class of functions of interest,  $\mathcal{F}$ , is the product of several functional classes. It is Glivenko-Cantelli if each of them is Glivenko-Cantelli.

Let  $\{t_h\}$ ,  $\{v_j\}$  and  $\{\bar{\beta}_m\}$  be the grid points of finite partitions of the intervals  $[0, \tau]$ ,  $[0, 1]^2$ , and  $\mathcal{B}$ , respectively. Let  $\{t_{h'}, t_h\}$ ,  $\{v_{j'}, v_j\}$  and  $\{\bar{\beta}_{m'}, \bar{\beta}_m\}$  be the grid points on the opposite ends of a hyper-cubic of the partitions such that  $0 \leq t_h - t_{h'} \leq \epsilon$ ,  $0 \leq v_j - v_{j'} \leq \epsilon$  and  $0 \leq \bar{\beta}_m - \bar{\beta}_{m'} \leq \epsilon$  for  $\epsilon > 0$ . Define the bracketing functions  $l_{h'j'm'} = S_{ki}^{(0)}(t_{h'}, v_{j'}, \bar{\beta}_{m'})$  and  $u_{hjm} = S_{ki}^{(0)}(t_h, v_j, \bar{\beta}_m)$ . Then for any  $f_{t,v,\bar{\beta}} \in \mathcal{F}$ , there is a bracket  $[l_{h'j'm'}, u_{hjm}]$  such that  $f_{t,v,\bar{\beta}} \in [l_{h'j'm'}, u_{hjm}]$ . Further,

$$\begin{split} \|u_{hjm} - l_{h'j'm'}\|_{P_{k},2} &\leq \|S_{ki}^{(0)}(t_{h}, v_{j}, \bar{\beta}_{m}) - S_{ki}^{(0)}(t_{h'}, v_{j'}, \bar{\beta}_{m'})\|_{P_{k},2} \\ &= \|Y_{ki}(t_{h})exp\{\bar{\beta}_{m}^{T}\tilde{Z}_{ki}(t_{h}, v_{j})\} - Y_{ki}(t_{h'})exp\{\bar{\beta}_{m'}^{T}\tilde{Z}_{ki}(t_{h'}, v_{j'})\}\|_{P_{k},2} \\ &\leq [C_{1}\|t_{h} - t_{h'}\| + C_{2}\|v_{j} - v_{j'}\| + C_{3}\|\bar{\beta}_{m} - \bar{\beta}_{m'}\|]^{1/2} \\ &\leq C\epsilon^{1/2}, \end{split}$$

 $\mathbf{2}$ 

where  $C_1$ ,  $C_2$ ,  $C_3$  and C are some positive constants. Hence, the bracketing number  $N_{[]}(\epsilon^{1/2}, \mathcal{F}, L_2(P_k))$  is of the polynomial order  $(1/\epsilon)^{4p+3}$ . Thus  $N_{[]}(\epsilon, \mathcal{F}, L_2(P_k))$  is of the order  $(1/\epsilon)^{2(4p+3)}$ . By the Glivenko–Cantelli Theorem (Theorem 19.4 of van der Vaart, 1998),  $S_k^{(0)}(t, v, \bar{\beta})$  converges in probability uniformly to  $s_k^{(0)}(t, v, \bar{\beta})$  for  $(t, v, \bar{\beta}) \in [0, \tau] \times [0, 1]^2 \times \mathcal{B}$ .

Proof of Theorem 2.1.

Let

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$$g_n(\bar{\beta}) = n^{-1}(l(\bar{\beta}) - l(\bar{\beta}_0)) \\= n^{-1} \sum_{k=1}^K \sum_{i=1}^{n_k} \int_0^\tau \int_0^1 \left[ (\bar{\beta} - \bar{\beta}_0)^T \tilde{Z}_{ki}(s, u) - \log\left\{\frac{S_k^{(0)}(s, u, \bar{\beta})}{S_k^{(0)}(s, u, \bar{\beta}_0)}\right\} \right] N_{ki}(ds, du).$$

By the condition (A.2), Lemma 1 and the uniform convergence of  $n_k^{-1} \sum_{i=1}^{n_k} N_{ki}(t, v) \xrightarrow{P} \int_0^t \int_0^v s_k^{(0)}(s, u, \bar{\beta}_0) \lambda_{0k}(s, u) \, ds du$  (Gilbert, Mckeague and Sun, 2004), we have  $\eta_n(\bar{\beta}) \xrightarrow{P} \eta(\bar{\beta})$ , where  $\eta(\bar{\beta}) = \sum_{k=1}^K p_k E \left( \int_0^\tau \int_0^1 \left[ (\bar{\beta} - \bar{\beta}_0)^T \tilde{Z}_{ki}(s, u) - \log \left\{ \frac{s_k^{(0)}(s, u, \bar{\beta})}{s_k^{(0)}(s, u, \bar{\beta}_0)} \right\} \right] \right) s_k^{(0)}(s, u, \bar{\beta}_0) \lambda_{0k}(s, u) \, ds du$ , uniformly in  $\bar{\beta} \in \mathcal{B}$ . Further,  $-\partial^2 \eta_n(\bar{\beta}) / \partial \bar{\beta}^2 = n^{-1} I(\bar{\beta})$  converges in probability to  $\Sigma(\bar{\beta})$  uniformly in  $\bar{\beta} \in \mathcal{B}$ . The limiting matrix function is a positive definite matrix for  $\bar{\beta} \in \mathcal{B}$  under the conditions (A.2) and (A.3). Hence,  $\eta_n(\bar{\beta})$  converges in probability to  $\eta(\bar{\beta})$ , we have  $\hat{\beta}$  converges in probability to  $\bar{\beta}_0$  as  $n \to \infty$ , see van der Vaart (1998).  $\Box$ 

Proof of Theorem 2.2.

Note that  $U(\hat{\beta}) - U(\bar{\beta}_0) = -I(\bar{\beta}^*)(\hat{\beta} - \bar{\beta}_0)$ , where  $\bar{\beta}^*$  is on the line segment between  $\hat{\beta}$  and  $\bar{\beta}_0$ . By the uniform convergence of  $n^{-1}I(\bar{\beta}) \xrightarrow{P} \Sigma(\bar{\beta})$  in probability in  $\bar{\beta} \in \mathcal{B}$  and the consistency of  $\hat{\beta}$  to  $\bar{\beta}_0$ , we have

$$n^{1/2}(\bar{\beta} - \bar{\beta}_0) = (I(\bar{\beta}^*)/n)^{-1} n^{-1/2} U(\bar{\beta}_0)$$
$$= (\Sigma(\bar{\beta}_0))^{-1} n^{-1/2} U(\bar{\beta}_0) + o_p(1).$$
(S.1)

It remains to show that  $n^{-1/2}U(\bar{\beta}_0) \xrightarrow{\mathcal{D}} N(0, \Sigma(\bar{\beta}_0))$ . Observe that

$$n^{-1/2}U(\bar{\beta}_0) = n^{-1/2} \sum_{k=1}^{K} \sum_{i=1}^{n_k} \int_0^{\tau} \int_0^1 \left[ \tilde{Z}_{ki}(t,u) - \frac{S_k^{(1)}(t,u,\bar{\beta}_0)}{S_k^{(0)}(t,u,\bar{\beta}_0)} \right] M_{ki}(dt,du).$$
(S.2)

By Lemma 1 of this Appendix and Lemma 2 of Gilbert, Mckeague and Sun (2008),

$$n^{-1/2}U(\bar{\beta}_0) = n^{-1/2} \sum_{k=1}^K \sum_{i=1}^{n_k} \int_0^\tau \int_0^1 \left[ \tilde{Z}_{ki}(t,u) - \frac{s_k^{(1)}(t,u,\bar{\beta})}{s_k^{(0)}(t,u,\bar{\beta})} \right] M_{ki}(dt,du) + o_p(1), \quad (S.3)$$

which converges in distribution to  $N(0, \Sigma(\bar{\beta}_0))$ . This completes the proof.

## Proof of Theorem 2.3.

Note that

$$\begin{split} &\Lambda_{0k}(t,v) - \Lambda_{0k}(t,v) \\ &= \int_0^t \int_0^v \left[ \frac{1}{n_k S_k^{(0)}(s,u,\hat{\beta})} - \frac{1}{n_k S_k^{(0)}(s,u,\bar{\beta}_0)} \right] N_{k\cdot}(ds,du) + \int_0^t \int_0^v \frac{M_{k\cdot}(ds,du)}{n_k S_k^{(0)}(s,u,\bar{\beta}_0)} + o_p(n_k^{-1/2}) \\ &= \int_0^t \int_0^v \frac{[S_k^{(1)}(s,u,\bar{\beta}_0)]^T(\bar{\beta}_0 - \hat{\beta})}{n_k S_k^{(0)}(s,u,\hat{\beta}) S_k^{(0)}(s,u,\bar{\beta}_0)} N_{k\cdot}(ds,du) + \int_0^t \int_0^v \frac{M_{k\cdot}(ds,du)}{n_k S_k^{(0)}(s,u,\bar{\beta}_0)} + o_p(n_k^{-1/2}) \\ &= \int_0^t \int_0^v \frac{[S_k^{(1)}(s,u,\bar{\beta}_0)]^T(\bar{\beta}_0 - \hat{\beta})}{S_k^{(0)}(s,u,\hat{\beta})} \lambda_{0k}(s,u) \, ds du + \int_0^t \int_0^v \frac{M_{k\cdot}(ds,du)}{n_k S_k^{(0)}(s,u,\bar{\beta}_0)} + o_p(n_k^{-1/2}), \quad (S.4) \end{split}$$

where  $M_{k.}(t,v) = \sum_{i=1}^{n_k} M_{ki}(t,v)$ . By Slusky Theorem, Lemma 2 of Gilbert, Mckeague and Sun (2008), Lemma A.2 of Sun et al. (2009), we have

$$\begin{split} &\sqrt{n}\{\hat{\Lambda}_{0k}(t,v) - \Lambda_{0k}(t,v)\} \\ &= -\sqrt{n}(\hat{\beta} - \bar{\beta}_0) \int_0^t \int_0^v \frac{[s_k^{(1)}(s,u,\bar{\beta}_0)]}{s_k^{(0)}(s,u,\bar{\beta}_0)} \lambda_{0k}(s,u) \, ds du + \sqrt{n} \int_0^t \int_0^v \frac{M_{k\cdot}(ds,du)}{n_k s_k^{(0)}(s,u,\bar{\beta}_0)} + o_p(1). \end{split}$$

Since  $M_{ki}(t, v)$ 's are mark-specific martingales, by (S.1) and (S.2), it is easy to check that  $\sqrt{n}(\hat{\beta} - \bar{\beta}_0)$  is asymptotically independent of the processes  $\sqrt{n} \int_0^t \int_0^v \frac{M_{k.}(ds,du)}{n_k s_k^{(0)}(s,u,\bar{\beta}_0)}$ ,  $k = 1, \ldots, K$ . Following the proof of Lemma A.2 of Sun et al. (2009), it can be shown that the latter are asymptotically independent mean-zero Guassian random fields, each of them has independent increments.  $\Box$  Proof of Theorem 3.1.

The cumulative martingale residuals are defined as

$$\hat{M}_{ki}(t,v) = \int_0^t \int_0^v N_{ki}(ds, du) - Y_{ki}(s) \exp(\hat{\beta}^T \tilde{Z}_{ki}(s, u)) \hat{\Lambda}_{0k}(ds, du),$$

where  $\hat{\Lambda}_{0k}(t,v) = \int_0^t \int_0^v \{n_k S_k^{(0)}(s,u,\hat{\beta})\}^{-1} N_{k.}(ds,du)$  is the estimator of the doubly cumulative baseline function  $\Lambda_{0k}(t,v) = \int_0^t \int_0^v \lambda_{0k}(s,u) ds du$ . It follows that

$$\hat{M}_{ki}(t,v) = M_{ki}(t,v) - \int_{0}^{t} \int_{0}^{v} Y_{ki}(s) \exp(\hat{\beta}^{T} \tilde{Z}_{ki}(s,u)) [\hat{\Lambda}_{0k}(ds,du) - \Lambda_{0k}(ds,du)] - \int_{0}^{t} \int_{0}^{v} Y_{ki}(s) [\exp(\hat{\beta}^{T} \tilde{Z}_{ki}(s,u)) - \exp(\bar{\beta}_{0}^{T} \tilde{Z}_{ki}(s,u)] \Lambda_{0k}(ds,du)].$$
(S.5)

By (S.4), we have

$$W_{k}(t,v,z) = n^{-1/2} \sum_{i=1}^{n_{k}} g_{k}(Z_{ki},z) M_{ki}(t,v) + n_{k} n^{-1/2} \int_{0}^{t} \int_{0}^{v} \frac{S_{kg}^{(0)}(s,u,z,\hat{\beta}) [S_{k}^{(1)}(s,u,\bar{\beta}_{0})]^{T}}{S_{k}^{(0)}(s,u,\bar{\beta}_{0})} \lambda_{0k}(s,u) \, ds du \, (\hat{\beta} - \bar{\beta}_{0}) - n^{-1/2} \int_{0}^{t} \int_{0}^{v} \frac{S_{kg}^{(0)}(s,u,z,\hat{\beta})}{S_{k}^{(0)}(s,u,\bar{\beta}_{0})} M_{k} \cdot (ds, du) - n_{k} n^{-1/2} \int_{0}^{t} \int_{0}^{v} [S_{kg}^{(1)}(s,u,z,\bar{\beta}_{0})]^{T} \lambda_{0k}(s,u) \, ds du \, (\hat{\beta} - \bar{\beta}_{0}) + o_{p}(1).$$
(S.6)

Replacing  $\hat{\beta}$  in  $S_{kg}^{(0)}(s, u, z, \hat{\beta})$  by  $\bar{\beta}_0$  results in a process that is equivalent to  $W_k(t, v, z)$  in probability by Slusky Theorem for the second term and by Lemma 2 of Gilbert, Mckeague and Sun (2008) for the third term of (S.6).

From the proof of Theorem 2, we have  $n^{1/2}(\hat{\beta} - \bar{\beta}_0) = (\Sigma(\bar{\beta}_0))^{-1}n^{-1/2}U(\bar{\beta}_0) + o_p(1)$ . The second term of (S.6) equals

$$(n_k/n) \left\{ \int_0^t \int_0^v \frac{[S_k^{(1)}(s, u, \bar{\beta}_0) \otimes S_{kg}^{(0)}(s, u, z, \bar{\beta})]^T}{S_k^{(0)}(s, u, \bar{\beta}_0)} \lambda_{0k}(s, u) \, ds du \right\} (\Sigma(\bar{\beta}_0))^{-1} n^{-1/2} U(\bar{\beta}_0) + o_p(1)$$

Similarly, the fourth term of (S.6) is equal to

$$(n_k/n) \left\{ \int_0^t \int_0^v [S_{kg}^{(1)}(s, u, z, \bar{\beta}_0)]^T \lambda_{0k}(s, u) \, ds du \right\} (\Sigma(\bar{\beta}_0))^{-1} n^{-1/2} U(\bar{\beta}_0) + o_p(1).$$

Bringing the above expressions into (S.6), we have

$$\begin{split} W_{k}(t,v,z) &= n^{-1/2} \sum_{i=1}^{n_{k}} \int_{0}^{t} \int_{0}^{v} \left[ g_{k}(Z_{ki},z) - \frac{S_{kg}^{(0)}(s,u,z,\bar{\beta}_{0})}{S_{k}^{(0)}(s,u,z,\bar{\beta}_{0})} \right] M_{ki}(ds,du) \\ &+ n_{k}/n \int_{0}^{t} \int_{0}^{v} \left( \frac{S_{k}^{(1)}(s,u,\bar{\beta}_{0}) \otimes S_{kg}^{(0)}(s,u,z,\bar{\beta}_{0})}{S_{k}^{(0)}(s,u,\bar{\beta}_{0})} - S_{kg}^{(1)}(s,u,z,\bar{\beta}_{0}) \right)^{T} \lambda_{0k}(s,u) \, dsdu \\ &(\Sigma(\bar{\beta}_{0}))^{-1} n^{-1/2} U(\bar{\beta}_{0}) + o_{p}(1) \\ &= n^{-1/2} \sum_{i=1}^{n_{k}} \int_{0}^{t} \int_{0}^{v} \left[ g_{k}(Z_{ki},z) - \frac{S_{kg}^{(0)}(s,u,z,\bar{\beta}_{0})}{S_{k}^{(0)}(s,u,\bar{\beta}_{0})} \right] M_{ki}(ds,du) \\ &+ (R_{k}(t,v,z))^{T} (\Sigma(\bar{\beta}_{0}))^{-1} n^{-1/2} \sum_{l=1}^{K} \sum_{i=1}^{n_{l}} \left\{ \int_{0}^{\tau} \int_{0}^{t} \left[ \tilde{Z}_{li}(s,u) - \frac{S_{l}^{(1)}(s,u,\bar{\beta}_{0})}{S_{l}^{(0)}(s,u,\bar{\beta}_{0})} \right] M_{li}(ds,du) \right\}^{T} \\ &+ o_{p}(1) \\ &= n^{-1/2} \sum_{l=1}^{K} \sum_{i=1}^{n_{l}} \int_{0}^{\tau} \int_{0}^{v} I(l=k)I(s\leqslant t) \left[ g_{l}(Z_{li},z) - \frac{S_{lg}^{(0)}(s,u,z,\bar{\beta}_{0})}{S_{l}^{(0)}(s,u,\bar{\beta}_{0})} \right] M_{li}(ds,du) \\ &+ (R_{k}(t,v,z))^{T} (\Sigma(\bar{\beta}_{0}))^{-1} n^{-1/2} \sum_{l=1}^{K} \sum_{i=1}^{n_{l}} \left\{ \int_{0}^{\tau} \int_{0}^{\tau} \left[ \tilde{Z}_{li}(s,u) - \frac{S_{l}^{(1)}(s,u,\bar{\beta}_{0})}{S_{l}^{(0)}(s,u,\bar{\beta}_{0})} \right] M_{li}(ds,du) \\ &+ (R_{k}(t,v,z))^{T} (\Sigma(\bar{\beta}_{0}))^{-1} n^{-1/2} \sum_{l=1}^{K} \sum_{i=1}^{n_{l}} \left\{ \int_{0}^{\tau} \int_{0}^{\tau} \left[ \tilde{Z}_{li}(s,u) - \frac{S_{l}^{(1)}(s,u,\bar{\beta}_{0})}{S_{l}^{(0)}(s,u,\bar{\beta}_{0})} \right] M_{li}(ds,du) \\ &+ (R_{k}(t,v,z))^{T} (\Sigma(\bar{\beta}_{0}))^{-1} n^{-1/2} \sum_{l=1}^{K} \sum_{i=1}^{n_{l}} \left\{ \int_{0}^{\tau} \int_{0}^{\tau} \left[ \tilde{Z}_{li}(s,u) - \frac{S_{l}^{(1)}(s,u,\bar{\beta}_{0})}{S_{l}^{(0)}(s,u,\bar{\beta}_{0})} \right] M_{li}(ds,du) \\ &+ (R_{k}(t,v,z))^{T} (\Sigma(\bar{\beta}_{0}))^{-1} n^{-1/2} \sum_{l=1}^{K} \sum_{i=1}^{n_{l}} \left\{ \int_{0}^{\tau} \int_{0}^{\tau} \left[ \tilde{Z}_{li}(s,u) - \frac{S_{l}^{(1)}(s,u,\bar{\beta}_{0})}{S_{l}^{(0)}(s,u,\bar{\beta}_{0})} \right] M_{li}(ds,du) \\ &+ (R_{k}(t,v,z))^{T} (\Sigma(\bar{\beta}_{0}))^{-1} n^{-1/2} \sum_{l=1}^{K} \sum_{i=1}^{n_{l}} \left\{ \int_{0}^{\tau} \int_{0}^{\tau} \left[ \tilde{Z}_{li}(s,u) - \frac{S_{l}^{(1)}(s,u,\bar{\beta}_{0})}{S_{l}^{(0)}(s,u,\bar{\beta}_{0})} \right] M_{li}(ds,du) \\ &+ (R_{k}(t,v,z))^{T} (\Sigma(\bar{\beta}_{0}))^{-1} n^{-1/2} \sum_{l=1}^{K} \sum_{i=1}^{n_{l}} \left\{ \int_{0}^{\tau} \int_{0}^{\tau} \int_{0}^{\tau} \left[ \tilde{Z}_{li}(s,u)$$

By the uniform convergence of  $S_k^{(0)}(s, u, \bar{\beta}_0)$ ,  $S_k^{(1)}(s, u, \bar{\beta}_0)$ ,  $S_{kg}^{(0)}(s, u, z, \bar{\beta}_0)$  and  $S_{kg}^{(1)}(s, u, z, \bar{\beta}_0)$  to  $s_k^{(0)}(s, u, \bar{\beta}_0)$ ,  $s_k^{(1)}(s, u, \bar{\beta}_0)$ ,  $s_{kg}^{(0)}(s, u, z, \bar{\beta}_0)$  and  $s_{kg}^{(1)}(s, u, z, \bar{\beta}_0)$  in  $(s, u) \in [0, \tau] \times [0, 1]^2$  in probability, respectively, and by the weak convergence of  $n_k^{-1/2} \sum_{i=1}^{n_k} M_{ki}(t, v)$  for  $k = 1, \ldots K$ , the terms  $S_k^{(0)}(s, u, \bar{\beta}_0)$ ,  $S_k^{(1)}(s, u, \bar{\beta}_0)$ ,  $S_{kg}^{(0)}(s, u, z, \bar{\beta}_0)$  and  $S_{kg}^{(1)}(s, u, z, \bar{\beta}_0)$  can be replaced by their expected values respectively. This completes the proof.  $\Box$