

# Supplementary Material for “Specification Testing for Transformation Models with an Application to Generalized Accelerated Failure-time Models”

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THIS SUPPLEMENTARY MATERIAL PROVIDES THE PROOFS OF ALL THE LEMMAS IN THE TEXT.

**Proof of Lemma B.1.** By Lemma 10.1 in Hoderlein et al. (2014) (HSWY),  $\hat{\beta}(y|v) - \beta(y|v) = \bar{\mathbf{S}}_b(v)^{-1} [\mathbf{V}_b(y;v) + \bar{\mathbf{B}}_b(y;v)] + O_P(\nu_{1b}^2 + \nu_{1b}b^{p+1}) = O_P(\nu_{1b} + b^{p+1})$  uniformly in  $(y, v) \in \mathcal{Y}_0 \times \mathcal{V}$ . The results follow from the fact that  $D_x \hat{F}_b(y|v) - D_x F(y|v) = e_1 [\hat{\beta}(y|x, z) - \beta(y|v)]/b$ . ■

**Proof of Lemma B.2.** The results follow from Lemma 10.5 in HSWY (2014) who prove the results based on standard arguments as used in Masry (1996), Hansen (2008), and Kong et al. (2010). ■

**Proof of Lemma B.3.** (a) Let  $\hat{q}(y;v) \equiv \hat{r}(y;v) - r(y;v)$ . Noting that  $\hat{f}_c(y|v)^{-1} = f(y|v)^{-1} - [\hat{f}_c(y|v) - f(y|v)]/f(y|v)^2 + R_1(y;v)$  where  $R_1(y;v) \equiv [\hat{f}_c(y|v) - f(y|v)]^2/[f(y|v)^2 \hat{f}_c(y|v)]$ , we have that for any  $(y, v) \in \mathcal{Y}_0 \times \mathcal{V}$ ,

$$\begin{aligned} \hat{q}(y;v) &= \frac{D_x \hat{F}_b(y|v)}{\hat{f}_c(y|v)} - \frac{D_x F(y|v)}{f(y|v)} = \frac{D_x \hat{F}_b(y|v) - D_x F(y|v)}{f(y|v)} + \left[ \frac{1}{\hat{f}_c(y|v)} - \frac{1}{f(y|v)} \right] D_x F(y|v) + R_2(y;v) \\ &= \frac{D_x \hat{F}_b(y|v) - D_x F(y|v)}{f(y|v)} - \frac{\hat{f}_c(y|v) - f(y|v)}{f(y|v)^2} D_x F(y|v) + R_1(y;v) D_x F(y|v) + R_2(y;v) \\ &\equiv \hat{q}_1(y;v) - \hat{q}_2(y;v) + R_1(y;v) D_x F(y|v) + R_2(y;v), \text{ say,} \end{aligned}$$

where  $R_2(y;v) \equiv [\hat{f}_c(y|v)^{-1} - f(y|v)^{-1}] [D_x \hat{F}_b(y|v) - D_x F(y|v)]$ . Using Lemmas B.1 and B.2, we can bound the last two terms in the last expression uniformly by  $O_p(\eta_{2c}(\eta_{1b} + \eta_{2c}))$ , where  $\eta_{1b} \equiv \nu_{1b}b^{-1} + b^p$  and  $\eta_{2c} \equiv \nu_{2c} + c^{p+1} + c^r$ . In addition, uniformly in  $(y, v) \in \mathcal{Y}_0 \times \mathcal{V}$ ,

$$\begin{aligned} \hat{q}_1(y;v) &= [D_x \hat{F}_b(y|v) - D_x F(y|v)]/f(y|v) \\ &= b^{-1} e_1 \bar{\mathbf{S}}_b(v)^{-1} [\mathbf{V}_b(y;v) + \bar{\mathbf{B}}_b(y;v)]/f(y|v) + O_P(\nu_{1b}^2 b^{-1} + \nu_{1b} b^p) \\ &= b^{-1} e_1 \bar{\mathbf{S}}_b(v)^{-1} \mathbf{V}_b(y;v)/f(y|v) + O_P(\nu_{1b}^2 b^{-1} + b^p), \end{aligned}$$

and

$$\begin{aligned} \hat{q}_2(y;v) &= D_x F(y|v) [\hat{f}_c(y|v) - f(y|v)]/f(y|v)^2 \\ &= D_x F(y|v) e'_2 \bar{\mathbf{S}}_c(v)^{-1} [\mathbf{V}_c^{(L)}(y;v) + \bar{\mathbf{B}}_c^{(L)}(y;v)]/f(y|v)^2 + O_P(\nu_{2c}^2 + \nu_{2c} c^{p+1}) \\ &= D_x F(y|v) e'_2 \bar{\mathbf{S}}_c(v)^{-1} \mathbf{V}_c^{(L)}(y;v)/f(y|v)^2 + O_P(\nu_{2c}^2 + c^{p+1} + c^r). \end{aligned}$$

It follows that uniformly in  $(y, v) \in \mathcal{Y}_0 \times \mathcal{V}$ ,

$$\begin{aligned} \hat{q}(y;v) &= b^{-1} e_1 \bar{\mathbf{S}}_b(v)^{-1} \mathbf{V}_b(y;v) f(y|v)^{-1} - D_x F(y|v) e'_2 \bar{\mathbf{S}}_c(v)^{-1} \mathbf{V}_c^{(L)}(y;v) f(y|v)^{-2} \\ &\quad + O_P(\nu_{1b}^2 b^{-1} + b^p + \nu_{2c}^2 + c^{p+1} + c^r + \eta_{2c}(\eta_{1b} + \eta_{2c})) \\ &= b^{-1} e_1 \bar{\mathbf{S}}_b(v)^{-1} \mathbf{V}_b(y;v) f(y|v)^{-1} - D_x F(y|v) e'_2 \bar{\mathbf{S}}_c(v)^{-1} \mathbf{V}_c^{(L)}(y;v) f(y|v)^{-2} + O_P(\nu_{bc}). \end{aligned}$$

(b) Write  $\hat{r}_0 - r_0 = \hat{r}_{01} + \hat{r}_{02}$ , where  $\hat{r}_{01} = \frac{1}{n^2} \sum_{i=1}^n \sum_{j=1}^n [\hat{r}(Y_i; X_j, Z_j) - r(Y_i; X_j, Z_j)]$  and  $\hat{r}_{02} = \frac{1}{n^2} \sum_{i=1}^n \sum_{j=1}^n [r(Y_i; X_j, Z_j) - r_0]$ . It is easy to show that  $\hat{r}_{02} = O_P(n^{-1/2})$  by the Chebyshev inequality. For  $\hat{r}_{01}$ , we have by (a) that  $\hat{r}_{01} = R_{1n} - R_{2n} + O_P(\nu_{bc})$ , where  $R_{1n} \equiv \frac{1}{n^2} \sum_{i=1}^n \sum_{j=1}^n b^{-1} e_1 \bar{\mathbf{S}}_b(V_j)^{-1} \mathbf{V}_b(Y_i; V_j) \times f_{ij}^{-1} \mathbf{1}_i$ ,  $R_{2n} \equiv \frac{1}{n^2} \sum_{i=1}^n \sum_{j=1}^n D_{xij} e_2' \bar{\mathbf{S}}_c(V_j)^{-1} \mathbf{V}_c^{(L)}(Y_i; V_j) f_{ij}^{-2} \mathbf{1}_i$ ,  $\mathbf{1}_i \equiv \mathbf{1}\{Y_i \in \mathcal{Y}_0\}$ ,  $f_{ij} \equiv f(Y_i|V_j)$ ,  $f_i \equiv f(Y_i|V_i)$  and  $D_{xij} \equiv D_x F(Y_i|V_j)$ . For  $R_{1n}$ , we have

$$\begin{aligned}
R_{1n} &= \frac{1}{n^3} \sum_{i=1}^n \sum_{j=1}^n \sum_{k=1}^n b^{-1} e_1 \bar{\mathbf{S}}_b(V_j)^{-1} \mu_b(V_k - V_j) K_b(V_k - V_j) \bar{\mathbf{I}}_{Y_i}(W_k) f_{ij}^{-1} \mathbf{1}_i \\
&= \frac{1}{n^3} \sum_{i=1}^n \sum_{j=1, j \neq i}^n \sum_{k=1, k \neq j, i}^n b^{-1} e_1 \bar{\mathbf{S}}_b(V_j)^{-1} \mu_b(V_k - V_j) K_b(V_k - V_j) \bar{\mathbf{I}}_{Y_i}(W_k) f_{ij}^{-1} \mathbf{1}_i \\
&\quad + \frac{1}{n^3} \sum_{i=1}^n \sum_{j=1, j \neq i}^n b^{-1} e_1 \bar{\mathbf{S}}_b(V_j)^{-1} \mu_b(V_i - V_j) K_b(V_i - V_j) \bar{\mathbf{I}}_{Y_i}(W_i) f_{ij}^{-1} \mathbf{1}_i \\
&\quad + \frac{1}{n^3} \sum_{i=1}^n \sum_{j=1, j \neq i}^n b^{-1} e_1 \bar{\mathbf{S}}_b(V_j)^{-1} \mu_b(0) K_b(0) \bar{\mathbf{I}}_{Y_i}(W_j) f_{ij}^{-1} \mathbf{1}_i \\
&\quad + \frac{1}{n^3} \sum_{i=1}^n \sum_{k=1, i}^n b^{-1} e_1 \bar{\mathbf{S}}_b(V_i)^{-1} \mu_b(V_k - V_i) K_b(V_k - V_i) \bar{\mathbf{I}}_{Y_i}(W_k) f_i^{-1} \mathbf{1}_i \\
&\quad + \frac{1}{n^3} \sum_{i=1}^n b^{-1} e_1 \bar{\mathbf{S}}_b(V_i)^{-1} \mu_b(0) K_b(0) \bar{\mathbf{I}}_{Y_i}(W_i) f_i^{-1} \mathbf{1}_i \\
&\equiv R_{1n,1} + R_{1n,2} + R_{1n,3} + R_{1n,4} + R_{1n,5}.
\end{aligned}$$

It is easy to show that  $R_{1n,5} = O_P(n^{-2}b^{-d-1})$ ,  $R_{1n,4} = O_P(n^{-3/2}b^{-1})$ ,  $R_{1n,3} = O_P(n^{-3/2}b^{-d-1})$ , and  $R_{1n,2} = O_P(n^{-1}b^{-1})$ . Noting that  $R_{1n,1}$  is a third-order  $U$ -statistic with  $E(R_{1n,1}) = 0$ , it is straightforward to show that  $E(R_{1n,1}^2) = O(n^{-1}b^{-2} + n^{-2}b^{-d-2})$ . Thus  $R_{1n,1} = O_P(n^{-1/2}b^{-1})$  and  $R_{1n} = O_P(n^{-1/2}b^{-1})$  as  $n^{-1}b^{-d} = o(1)$ . By the same token, we can show that  $R_{2n} = O_P(n^{-1/2})$ . It follows that  $\hat{r}_0 - r_0 = O_P(\nu_{bc} + n^{-1/2}b^{-1})$ .

(c) Write  $\hat{r}_2(y) - r_2(y) = \hat{r}_{21}(y) + \hat{r}_{22}(y)$ , where  $\hat{r}_{21}(y) = \frac{1}{n} \sum_{i=1}^n [\hat{r}(y; V_i) - r(y; V_i)]$  and  $\hat{r}_{22}(y) = \frac{1}{n} \sum_{i=1}^n [r(y; V_i) - r_2(y)]$ . By standard chaining arguments and the exponential inequality, we can show that  $\sup_{y \in \mathcal{Y}_0} \|\hat{r}_{22}(y)\| = O(n^{-1/2} \sqrt{\ln n})$ . By (a),  $\hat{r}_{21}(y) = \bar{r}_{21}(y) + O_P(\nu_{bc})$  uniformly in  $y \in \mathcal{Y}_0$ , where  $\bar{r}_{21}(y) \equiv \bar{r}_{21,1}(y) - \bar{r}_{21,2}(y)$ ,  $\bar{r}_{21,1}(y) \equiv \frac{1}{n} \sum_{i=1}^n b^{-1} e_1 \bar{\mathbf{S}}_b(V_i)^{-1} \mathbf{V}_b(y; V_i) f(y|V_i)^{-1} \mathbf{1}\{y \in \mathcal{Y}_0\}$ , and  $\bar{r}_{21,2}(y) \equiv \frac{1}{n} \sum_{i=1}^n f(y|V_i)^{-2} D_x F(y|V_i) \times e_2' \bar{\mathbf{S}}_c(V_i)^{-1} \mathbf{V}_c^{(L)}(y; V_i) \mathbf{1}\{y \in \mathcal{Y}_0\}$ . Now write  $\bar{r}_{21,1}(y)$  as the summation of a first order  $U$ -statistic and a second order  $U$ -statistic:  $\bar{r}_{21,1}(y) = \bar{r}_{21,11}(y) + \bar{r}_{21,12}(y)$ , where

$$\begin{aligned}
\bar{r}_{21,11}(y) &\equiv \frac{1}{n^2} \sum_{i=1}^n b^{-1} e_1 \bar{\mathbf{S}}_b(V_i)^{-1} \mu_b(0) K_b(0) \bar{\mathbf{I}}_y(W_i) f(y|V_i)^{-1} \mathbf{1}\{y \in \mathcal{Y}_0\}, \text{ and} \\
\bar{r}_{21,12}(y) &\equiv \frac{1}{n^2} \sum_{i=1}^n \sum_{j=1, j \neq i}^n b^{-1} e_1 \bar{\mathbf{S}}_b(V_i)^{-1} \mu_b(V_j - V_i) K_b(V_j - V_i) \bar{\mathbf{I}}_y(W_j) f(y|V_i)^{-1} \mathbf{1}\{y \in \mathcal{Y}_0\}.
\end{aligned}$$

By the exponential inequality, we can show that  $\sup_{y \in \mathcal{Y}_0} \|\bar{r}_{21,11}(y)\| = O(n^{-3/2}b^{-d-1} \sqrt{\ln n})$ . For  $\bar{r}_{21,12}(y)$ , one can follow the proof of (A.10) in Gozalo and Linton (2001) and show that  $\sup_{y \in \mathcal{Y}_0} \|\bar{r}_{21,12}(y)\| = O(n^{-1/2}b^{-1} \sqrt{\ln n})$ .<sup>1</sup> Hence  $\sup_{y \in \mathcal{Y}_0} \|\bar{r}_{21,1}(y)\| = O(n^{-1/2}b^{-1} \sqrt{\ln n})$ . Similarly,  $\sup_{y \in \mathcal{Y}_0} \|\bar{r}_{21,2}(y)\| = O(n^{-1/2} \sqrt{\ln n})$ . Thus  $\sup_{y \in \mathcal{Y}_0} \|\hat{r}_2(y) - r_2(y)\| = O_P(\nu_{bc} + n^{-1/2}b^{-1} \sqrt{\ln n})$ . ■

<sup>1</sup>If we ignore the boundary points, we can write  $\bar{\mathbf{S}}_b(v) = f(v) \mathbb{S} + b \mathbb{V}(v) + o(b)$  uniformly in  $v$  in the interior of  $\mathcal{V}$ , where

**Proof of Lemma B.4.** Recall  $\mathbf{1}_i \equiv \mathbf{1}\{Y_i \in \mathcal{Y}_0\}$ . Let  $f_i \equiv f(Y_i|V_i)$  and  $D_{xi} \equiv D_x F(Y_i|X_i, Z_i)$ . By Lemma B.3(a),  $\hat{r}_i - r_i = b^{-1}e_1 \bar{\mathbf{S}}_b(V_i)^{-1} \mathbf{V}_b(Y_i; V_i) f_i^{-1} \mathbf{1}_i - D_{xi} e_2' \bar{\mathbf{S}}_c(V_i)^{-1} \mathbf{V}_c^{(L)}(Y_i; V_i) f_i^{-2} \mathbf{1}_i + O_P(\nu_{bc})$ . It follows that

$$\begin{aligned} \Gamma_{2n,1} &= b^{\frac{d}{2}+2} \sum_{i=1}^n \|(\hat{r}_i - r_i) \circ r_0\|^2 a_i \\ &= b^{\frac{d}{2}+2} \sum_{i=1}^n \left\| \left( b^{-1}e_1 \bar{\mathbf{S}}_b(V_i)^{-1} \mathbf{V}_b(Y_i; V_i) f_i^{-1} - D_{xi} e_2' \bar{\mathbf{S}}_c(V_i)^{-1} \mathbf{V}_c^{(L)}(Y_i; V_i) f_i^{-2} \right) \circ r_0 \right\|^2 a_i \\ &\quad + nb^{\frac{d}{2}+2} O_P(\nu_{bc}^2 + \nu_{bc}(\nu_{1b}b^{-1} + b^p + \nu_{2c} + c^{p+1} + c^r)) \\ &= \bar{\Gamma}_{2n,1} + o_P(1), \end{aligned}$$

where  $\bar{\Gamma}_{2n,1} \equiv b^{\frac{d}{2}+2} \sum_{i=1}^n \left\| \left( b^{-1}e_1 \bar{\mathbf{S}}_b(V_i)^{-1} \mathbf{V}_b(Y_i; V_i) f_i^{-1} - D_{xi} e_2' \bar{\mathbf{S}}_c(V_i)^{-1} \mathbf{V}_c^{(L)}(Y_i; V_i) f_i^{-2} \right) \circ r_0 \right\|^2 a_i$ , and we use the fact that  $\mathbf{1}_i a_i = a_i$  as  $a(y; v)$  has compact support  $\mathcal{Y}_0 \times \mathcal{V}_0$ . Let  $\zeta_k(w) \equiv \zeta_k(y; v)$  be as defined in Section 3.2. Then

$$\bar{\Gamma}_{2n,1} = b^{\frac{d}{2}+2} \sum_{i=1}^n \left\| n^{-1} \sum_{k=1}^n \zeta_k(W_i) \circ r_0 \right\|^2 a_i = n^{-2} b^{\frac{d}{2}+2} \sum_{i_1=1}^n \sum_{i_2=1}^n \sum_{i_3=1}^n \zeta(W_{i_1}, W_{i_2}, W_{i_3}),$$

where  $\zeta(W_{i_1}, W_{i_2}, W_{i_3}) \equiv (\zeta_{i_2}(W_{i_1}) \circ r_0)' (\zeta_{i_3}(W_{i_1}) \circ r_0) a_{i_1}$ . Let  $\varphi(w_{i_1}, w_{i_2}) \equiv E[\zeta(W_1, w_{i_1}, w_{i_2})]$ , and  $\bar{\zeta}(w_{i_1}, w_{i_2}, w_{i_3}) \equiv \zeta(w_{i_1}, w_{i_2}, w_{i_3}) - \varphi(w_{i_2}, w_{i_3})$ . We can decompose  $\bar{\Gamma}_{2n,1}$  as  $\bar{\Gamma}_{2n,1} = \bar{\Gamma}_{2n,11} + \bar{\Gamma}_{2n,12}$ , where

$$\bar{\Gamma}_{2n,11} = n^{-1} b^{\frac{d}{2}+2} \sum_{i_1=1}^n \sum_{i_2=1}^n \varphi(W_{i_1}, W_{i_2}) \text{ and } \bar{\Gamma}_{2n,12} = n^{-2} b^{\frac{d}{2}+2} \sum_{i_1=1}^n \sum_{i_2=1}^n \sum_{i_3=1}^n \bar{\zeta}(W_{i_1}, W_{i_2}, W_{i_3}).$$

Consider  $\bar{\Gamma}_{2n,12}$  first. Write  $E(\bar{\Gamma}_{2n,12}^2) = n^{-4} b^{d+4} \sum_{i_1, \dots, i_6} E[\bar{\zeta}(W_{i_1}, W_{i_2}, W_{i_3}) \bar{\zeta}(W_{i_4}, W_{i_5}, W_{i_6})]$ . Noting that  $E[\bar{\zeta}(W_{i_1}, w_{i_2}, w_{i_3})] = E[\bar{\zeta}(w_{i_1}, W_{i_2}, w_{i_3})] = E[\bar{\zeta}(w_{i_1}, w_{i_2}, W_{i_3})] = 0$ ,  $E[\bar{\zeta}(W_{i_1}, W_{i_2}, W_{i_3}) \bar{\zeta}(W_{i_4}, W_{i_5}, W_{i_6})] = 0$  if there are more than three distinct elements in  $\{i_1, \dots, i_6\}$ . With this, it is easy to show that  $E(\bar{\Gamma}_{2n,12}^2) = O(n^{-1} b^{d+4} (b^{-4-3d} + c^{-3(d+1)})) = o(1)$ . Hence  $\bar{\Gamma}_{2n,12} = o_P(1)$  by the Chebyshev inequality.

For  $\bar{\Gamma}_{2n,11}$ , we have  $\bar{\Gamma}_{2n,11} = n^{-1} b^{\frac{d}{2}+2} \sum_{i=1}^n \varphi(W_i, W_i) + 2n^{-1} b^{\frac{d}{2}+2} \sum_{1 \leq i < j \leq n} \varphi(W_i, W_j) \equiv \mathbb{B}_{1n} + \mathbb{V}_{1n}$ , say, where  $\varphi(W_i, W_j) = \int \zeta(w, W_i, W_j) dF(w) = \int (\zeta_i(w) \circ r_0)' (\zeta_j(w) \circ r_0) a(w) dF(w)$ , and  $\mathbb{B}_{1n}$  and  $\mathbb{V}_{1n}$  contribute to the asymptotic bias and variance of  $\bar{\Gamma}_{2n,11}$ , respectively. Note that as  $\mathbb{V}_{1n}$  is a second-order degenerate  $U$ -statistic, we can easily verify that all the conditions of Theorem 1 of Hall (1984) are satisfied and a central limit theorem applies to it:  $\mathbb{V}_{1n} \xrightarrow{d} N(0, \sigma_0^2)$ , where  $\sigma_0^2 = \lim_{n \rightarrow \infty} \sigma_n^2$  and  $\sigma_n^2 = 2b^{d+4} E[\varphi(W_1, W_2)]^2$ . Thus  $\Gamma_{2n,11} - \mathbb{B}_{1n} \xrightarrow{d} N(0, \sigma_0^2)$ .

Lastly, noting that  $E|\mathbb{B}_{1n}| = b^{\frac{d}{2}+2} O(b^{-d-2} + c^{-d-1})$ , we have  $\mathbb{B}_{1n} = O_P(b^{\frac{d}{2}+2}(b^{-d-2} + c^{-d-1}))$  by Markov inequality. ■

**Proof of Lemma B.5.** (a) Note that  $\Gamma_{2n,2} \leq nb^{\frac{d}{2}+2} \|\hat{r}_0 - r_0\|^2 \bar{\Gamma}_{2n,2}$  where  $\bar{\Gamma}_{2n,2} \equiv n^{-1} \sum_{i=1}^n \|r_i\|^2 a_i$ . By Assumptions C.2(ii) and C.3(i), the compact support of  $a$ , and Markov inequality,  $\bar{\Gamma}_{2n,2} = O_P(1)$ . Using this and Lemma B.3(b) we have  $\Gamma_{2n,2} = nb^{\frac{d}{2}+2} O_P(\nu_{bc}^2 + n^{-1} b^{-2}) O_P(1) = o_P(1)$ .

$\mathbb{S}$  and  $\mathbb{V}$  are defined as  $M$  and  $V$  in Li et al. (2003, p. 617). Following the proof of Lemma A.3 in their paper, one can show that  $\bar{r}_{21,12}(y) = O(n^{-1/2})$  elementwise by using the degeneracy of the second order  $U$ -statistic defined analogously to  $\bar{r}_{21,12}(y)$  but with  $\bar{\mathbf{S}}_b(V_i)^{-1}$  replaced by its leading term  $f(V_i)^{-1} \mathbb{S}^{-1}$ . But their argument breaks down when  $v$  takes values on the boundary of  $\mathcal{V}$ .

<sup>2</sup>Write  $\bar{\Gamma}_{2n,1} = b^{\frac{d}{2}+2} \sum_{i=1}^n \|(\xi_{1ni} - \xi_{2ni}) \circ r_0\|^2 a_i$ , where  $\xi_{1ni} \equiv b^{-1}e_1 \bar{\mathbf{S}}_b(V_i)^{-1} \mathbf{V}_b(Y_i; V_i) f_i^{-1}$  and  $\xi_{2ni} \equiv D_{xi} e_2' \bar{\mathbf{S}}_c(V_i)^{-1} \mathbf{V}_c^{(L)}(Y_i; V_i) f_i^{-2}$ . By straightforward moment calculations, we can show that  $\xi_{1ni}$  contributes to both the asymptotic bias and variance of the test statistic whereas  $\xi_{2ni}$  only contributes to the asymptotic bias.

(b) Noting that  $\hat{r}_1(x) - r_1(x) = \frac{1}{n} \sum_{i=1}^n [\hat{r}(Y_i; x, Z_i) - r(Y_i; x, Z_i)] + \frac{1}{n} \sum_{i=1}^n [r(Y_i; x, Z_i) - r_1(x)]$ , we have  $b^{\frac{d}{2}+2} \sum_{i=1}^n \|\hat{r}_1(X_i) - r_1(X_i)\| \circ r_{2i}\|^2 a_i = R_{3n} + R_{4n} + 2R_{5n}$ , where

$$\begin{aligned} R_{3n} &\equiv \frac{b^{\frac{d}{2}+2}}{n^2} \sum_{i=1}^n \left\| \sum_{j=1}^n [\hat{r}(Y_j; X_i, Z_j) - r(Y_j; X_i, Z_j)] \circ r_{2i} \right\|^2 a_i, \\ R_{4n} &\equiv \frac{b^{\frac{d}{2}+2}}{n^2} \sum_{i=1}^n \left\| \sum_{j=1}^n [r(Y_j; X_i, Z_j) - r(X_i)] \circ r_{2i} \right\|^2 a_i, \text{ and} \\ R_{5n} &\equiv \frac{b^{\frac{d}{2}+2}}{n^2} \sum_{i=1}^n \left\| \sum_{j=1}^n [\hat{r}(Y_j; X_i, Z_j) - r(Y_j; X_i, Z_j)] \circ r_{2i} \right\| \left\| \sum_{k=1}^n [r(Y_k; X_i, Z_k) - r(X_i)] \circ r_{2i} \right\| a_i. \end{aligned}$$

By Lemma B.3(a) we can readily show that  $R_{3n} = \mathbb{B}_{2n} + o_P(1)$ . We further decompose  $\mathbb{B}_{2n}$  as  $\mathbb{B}_{2n} = \mathbb{B}_{2n,1} + \mathbb{B}_{2n,2}$ , where

$$\begin{aligned} \mathbb{B}_{2n,1} &\equiv \frac{b^{\frac{d}{2}+2}}{n^4} \sum_{i_1=1}^n \sum_{i_2=1}^n \sum_{i_3=1}^n \sum_{i_4=1}^n (\zeta_{i_3}(Y_{i_2}; X_{i_1}, Z_{i_2}) \circ r_{2i_1})' (\zeta_{i_3}(Y_{i_4}; X_{i_1}, Z_{i_4}) \circ r_{2i_1}) a_{i_1} \text{ and} \\ \mathbb{B}_{2n,2} &\equiv \frac{b^{\frac{d}{2}+2}}{n^4} \sum_{i_1=1}^n \sum_{i_2=1}^n \sum_{i_3=1}^n \sum_{i_4=1}^n \sum_{i_5=1, i_5 \neq i_3}^n (\zeta_{i_3}(Y_{i_2}; X_{i_1}, Z_{i_2}) \circ r_{2i_1})' (\zeta_{i_5}(Y_{i_4}; X_{i_1}, Z_{i_4}) \circ r_{2i_1}) a_{i_1}. \end{aligned}$$

By direct moment calculations and the Chebyshev inequality, we can show that  $\mathbb{B}_{2n,2} = O_P(b^{\frac{d_z}{2}} + b^{\frac{d}{2}+2} c^{-\frac{d_x}{2}})$  which is  $o_P(1)$  under Assumption A.7 if  $d_z > 0$ , and that

$$\mathbb{B}_{2n,1} = \bar{\mathbb{B}}_{2n} = \frac{1}{n^4} b^{\frac{d}{2}+2} \sum_{i_1=1}^n \sum_{i_3=1}^n \left\| \sum_{i_2=1}^n \zeta_{i_3}(Y_{i_2}; X_{i_1}, Z_{i_2}) \circ r_{2i_1} \right\|^2 a_{i_1} = O_P(b^{\frac{d}{2}+2} (b^{-d_x-2} + c^{-d_x})).$$

It follows that  $R_{3n} = \bar{\mathbb{B}}_{2n} + o_P(1)$  if  $d_z > 0$ . By Markov inequality,  $R_{4n} = O_P(b^{\frac{d}{2}+2}) = o_P(1)$ . By Cauchy-Schwarz inequality,  $R_{5n} \leq \{R_{3n}R_{4n}\}^{1/2} = O_P(\{[b^{\frac{d}{2}+2}(b^{-d_x-2} + c^{-d_x}) + 1]b^{\frac{d}{2}+2}\}^{1/2}) = o_P(1)$ . This completes the proof of part (b).

(c) Noting that  $\hat{r}_2(y) - r_2(y) = \frac{1}{n} \sum_{i=1}^n [\hat{r}(y; X_i, Z_i) - r(y; X_i, Z_i)] + \frac{1}{n} \sum_{i=1}^n [r(y; X_i, Z_i) - r_2(y)]$ , by Cauchy-Schwarz inequality we have  $\Gamma_{2n,4} \leq 2R_{6n} + 2R_{7n}$ , where  $R_{6n} \equiv \frac{b^{\frac{d}{2}+2}}{n^2} \sum_{i=1}^n \|\sum_{j=1}^n r_{1i} \circ [\hat{r}(Y_i; X_j, Z_j) - r(Y_i; X_j, Z_j)]\|^2 a_i$  and  $R_{7n} \equiv \frac{b^{\frac{d}{2}+2}}{n^2} \sum_{i=1}^n \|\sum_{j=1}^n r_{1i} \circ [r(Y_i; X_j, Z_j) - r_2(Y_i)]\|^2 a_i$ . By Markov inequality  $R_{7n} = O_P(b^{\frac{d}{2}+2}) = o_P(1)$ . For  $R_{6n}$  we can first apply Lemma B.3 to show that  $R_{6n} = \bar{R}_{6n} + o_P(1)$ , where

$$\bar{R}_{6n} \equiv \frac{b^{\frac{d}{2}+2}}{n^2} \sum_{i=1}^n \left\| \sum_{j=1}^n r_{1i} \circ \left[ b^{-1} e_1 \bar{\mathbf{S}}_b(V_j)^{-1} \mathbf{V}_b(Y_i; V_j) f_{ij}^{-1} - f_{ij}^{-2} D_{xij} e_2' \bar{\mathbf{S}}_c(X_j, Z_j)^{-1} \mathbf{V}_c^{(L)}(Y_i; V_j) \right] \right\|^2 a_i,$$

$f_{ij} \equiv f(Y_i|V_j)$ , and  $D_{xij} \equiv D_x F(Y_i|V_j)$ . Observe that  $\bar{R}_{6n} \leq 2\bar{R}_{6n,1} + 2\bar{R}_{6n,2}$ , where

$$\begin{aligned} \bar{R}_{6n,1} &= \frac{b^{\frac{d}{2}+2}}{n^2} \sum_{i=1}^n \left\| \sum_{j=1}^n r_{1i} \circ \left( b^{-1} e_1 \bar{\mathbf{S}}_b(V_j)^{-1} \mathbf{V}_b(Y_i; V_j) f_{ij}^{-1} \right) \right\|^2 a_i, \text{ and} \\ \bar{R}_{6n,2} &= \frac{b^{\frac{d}{2}+2}}{n^2} \sum_{i=1}^n \left\| \sum_{j=1}^n r_{1i} \circ \left( f_{ij}^{-2} D_{xij} e_2' \bar{\mathbf{S}}_c(V_j)^{-1} \mathbf{V}_c^{(L)}(Y_i; V_j) \right) \right\|^2 a_i. \end{aligned}$$

By straightforward but tedious moment calculations, we can show that

$$\begin{aligned} E(\bar{R}_{6n,1}) &= \frac{b^{\frac{d}{2}}}{n^4} \sum_{i=1}^n E \left\| \sum_{j=1}^n \sum_{k=1}^n r_{1i} \circ \left( e_1 \bar{\mathbf{S}}_b(V_j)^{-1} \mu_b(V_k - V_j) K_b(V_k - V_j) \bar{\mathbf{I}}_{Y_i}(W_k) f_{ij}^{-1} \right) \right\|^2 a_i \\ &= O\left(b^{\frac{d}{2}} + n^{-1}b^{-\frac{d}{2}} + n^{-2}b^{-\frac{3d}{2}}\right) = o(1). \end{aligned}$$

Similarly,  $E(\bar{R}_{6n,2}) = b^{\frac{d}{2}+2}O(1 + n^{-1}c^{-(d+1)} + n^{-2}c^{-2(d+1)}) = o(1)$ . Then  $\bar{R}_{6n} = o_P(1)$  by Markov inequality. It follows that  $R_{6n} = o_P(1)$  and  $\Gamma_{2n,4} = o_P(1)$ . ■

**Proof of Lemma B.6.** (a) Noting that  $\Gamma_{2n,5} = 2b^{\frac{d}{2}+2}((\hat{r}_0 - r_0) \circ r_0)' \bar{\Gamma}_{2n,5}$  where  $\bar{\Gamma}_{2n,5} = \sum_{i=1}^n ((\hat{r}_i - r_i) \circ r_i) a_i$ . By Lemma B.3(a), we can show that  $\bar{\Gamma}_{2n,5} = \bar{\Gamma}_{2n,51} + o_P(n^{1/2}b^{-(\frac{d}{2}+1)})$ , where

$$\begin{aligned} \bar{\Gamma}_{2n,51} &= \sum_{i=1}^n \left[ \left( b^{-1} e_1 \bar{\mathbf{S}}_b(V_i)^{-1} \mathbf{V}_b(Y_i; V_i) f_i^{-1} - D_{x_i} e_2' \bar{\mathbf{S}}_c(V_i)^{-1} \mathbf{V}_c^{(L)}(Y_i; V_i) f_i^{-2} \right) \circ r_i \right] a_i \\ &= \frac{1}{n} \sum_{i=1}^n \left[ \sum_{j=1}^n \zeta_{1j}(Y_i; V_i) \circ r_i \right] a_i - \frac{1}{n} \sum_{i=1}^n \left[ \sum_{j=1}^n \zeta_{2j}(Y_i; V_i) \circ r_i \right] a_i \equiv R_{8n} + R_{9n}. \end{aligned}$$

In view of  $R_{8n} = \frac{1}{n} \sum_{i=1}^n \left[ \sum_{j=1}^n [b^{-1} e_1 \bar{\mathbf{S}}_b(V_i)^{-1} \mu_b(V_j - V_i) K_b(V_j - V_i) \bar{\mathbf{I}}_{Y_i}(W_j) f_i^{-1}] \circ r_i \right] a_i$ , it is easy to show that  $E\|R_{8n}\|^2 = O(nb^{-2} + b^{-2d-2})$ . Thus  $R_{8n} = O_P(b^{-d-1} + n^{1/2}b^{-1})$ . Similarly,  $R_{9n} = O_P(c^{-d-1} + n^{1/2})$ . It follows that  $\bar{\Gamma}_{2n,51} = O_P(b^{-d-1} + n^{1/2}b^{-1} + c^{-d-1})$ . Then by Lemma B.3(b) and the fact that  $\nu_{bc} = O_P(n^{-1/2}b^{-1})$ ,  $\Gamma_{2n,5} = b^{\frac{d}{2}+2}O_P(n^{-1/2}b^{-1}) \left[ O_P(b^{-d-1} + n^{1/2}b^{-1} + c^{-d-1}) + o_P(n^{1/2}b^{-(\frac{d}{2}+1)}) \right] = o_P(1)$ .

(b) Write  $\Gamma_{2n,6} = -2r_0' \bar{\Gamma}_{2n,6}$  where  $\bar{\Gamma}_{2n,6} \equiv b^{\frac{d}{2}+2} \sum_{i=1}^n ((\hat{r}_i - r_i) \circ (\hat{r}_{1i} - r_{1i}) \circ r_{2i}) a_i$ . Then  $\bar{\Gamma}_{2n,6} = R_{10n} + R_{11n}$ , where  $R_{10n} \equiv n^{-1}b^{\frac{d}{2}+2} \sum_{i=1}^n \sum_{j=1}^n ((\hat{r}_i - r_i) \circ [\hat{r}(Y_j; X_i, Z_j) - r(Y_j; X_i, Z_j)] \circ r_{2i}) a_i$  and  $R_{11n} \equiv n^{-1}b^{\frac{d}{2}+2} \sum_{i=1}^n \sum_{j=1}^n ((\hat{r}_i - r_i) \circ [r(Y_j; X_i, Z_j) - r_1(X_i)] \circ r_{2i}) a_i$ . Using Lemma B.3, we can show that  $R_{10n} = \bar{R}_{10n} + o_P(1)$ , where

$$\begin{aligned} \bar{R}_{10n} &= n^{-1}b^{\frac{d}{2}+2} \sum_{i=1}^n \sum_{j=1}^n \left\{ \left[ b^{-1} e_1 \bar{\mathbf{S}}_b(V_i)^{-1} \mathbf{V}_b(Y_i; V_i) f_i^{-1} - D_{x_i} e_2' \bar{\mathbf{S}}_c(V_i)^{-1} \mathbf{V}_c^{(L)}(Y_i; V_i) f_i^{-2} \right] \right. \\ &\quad \left. \circ \left[ b^{-1} e_1 \bar{\mathbf{S}}_b(X_i, Z_j)^{-1} \mathbf{V}_b(Y_j; X_i, Z_j) f_{ji}^{-1} - D_{x_i} e_2' \bar{\mathbf{S}}_c(X_i, Z_j)^{-1} \mathbf{V}_c^{(L)}(Y_j; X_i, Z_j) f_{ji}^{-2} \right] \circ r_{2i} \right\} a_i \mathbf{1}_i \mathbf{1}_j \\ &= n^{-3}b^{\frac{d}{2}+2} \sum_{i=1}^n \left[ \sum_{l=1}^n \zeta_l(W_i) \circ \sum_{j=1}^n \sum_{k=1}^n \zeta_k(Y_j; X_i, Z_j) \circ r_2(Y_i) \right] a_i. \end{aligned}$$

Noting that  $E\|\bar{R}_{10n}\|^2 = O(b^{d+4}(b^{-2d_x-4} + c^{-2d_x}))$ , we have  $\bar{R}_{10n} = O_P(b^{(d_z-d_x)/2} + b^{\frac{d}{2}+2}c^{-d_x})$  which is  $o_P(1)$  under Assumption A.7 if  $d_z > d_x$  and otherwise not. Hence  $r_0' \bar{R}_{10n} = \mathbb{B}_{3n} + o_P(1)$  and  $\mathbb{B}_{3n} = O_P(b^{(d_z-d_x)/2} + b^{\frac{d}{2}+2}c^{-d_x})$ . For  $R_{11n}$ , we apply Cauchy-Schwarz inequality to obtain  $R_{11n} \leq \{\alpha_{1n} \alpha_{2n}\}^{1/2}$ , where  $\alpha_{1n} \equiv b^{\frac{d}{2}+2} \sum_{i=1}^n \|(\hat{r}_i - r_i) \circ r_{2i}\|^2 a_i$  and  $\alpha_{2n} \equiv n^{-1}b^{\frac{d}{2}+2} \sum_{i=1}^n \left\| \sum_{j=1}^n [r(Y_j; X_i, Z_j) - r_1(X_i)] \right\|^2 a_i$ . Analogously to the determination of the probability order of  $\Gamma_{2n,1}$ , we can show that  $\alpha_{1n} = O_P(b^{\frac{d}{2}+2}(b^{-d-2} + c^{-d-1}))$ . Next,  $\alpha_{2n} = O_P(b^{\frac{d}{2}+2})$  by Markov inequality. It follows that  $R_{11n} = O_P(b^{-\frac{d}{4}} + b^{\frac{d}{4}+1}c^{-\frac{d+1}{2}})O_P(b^{\frac{d}{4}+1}) = O_P(b + b^{\frac{d}{2}+2}c^{-\frac{d+1}{2}}) = o_P(1)$  and  $\Gamma_{2n,6} = -2\mathbb{B}_{3n} + o_P(1)$ .

(c) Write  $\Gamma_{2n,7} = -2r_0' \bar{\Gamma}_{2n,7}$  where  $\bar{\Gamma}_{2n,7} \equiv b^{\frac{d}{2}+2} \sum_{i=1}^n [(\hat{r}_i - r_i) \circ r_{1i} \circ (\hat{r}_{2i} - r_{2i})] a_i$ . We further decompose  $\bar{\Gamma}_{2n,7}$  as  $\bar{\Gamma}_{2n,7} = R_{12n} + R_{13n}$ , where  $R_{12n} \equiv n^{-1}b^{\frac{d}{2}+2} \sum_{i=1}^n \sum_{j=1}^n \{(\hat{r}_i - r_i) \circ r_{1i} \circ [\hat{r}(Y_i; V_j) - r(Y_i; V_j)]\} a_i$

and  $R_{13n} \equiv n^{-1}b^{\frac{d}{2}+2} \sum_{i=1}^n \sum_{j=1}^n \{(\hat{r}_i - r_i) \circ r_{1i} \circ [r(Y_i; V_j) - r_2(Y_i)]\} a_i$ . Following the analysis of  $R_{10}$  and  $R_{11n}$ , we can readily show that  $R_{sn} = o_P(1)$  for  $s = 12, 13$ . It follows that  $\Gamma_{2n,7} = o_P(1)$ .

(d) Noting that  $\Gamma_{2n,8} = 2b^{\frac{d}{2}+2}(\hat{r}_0 - r_0)' \bar{\Gamma}_{2n,8}$  where  $\bar{\Gamma}_{2n,8} \equiv \sum_{i=1}^n (r_i \circ (\hat{r}_{1i} - r_{1i}) \circ r_{2i}) a_i$ . Then  $\bar{\Gamma}_{2n,8} = R_{14n} + R_{15n}$ , where  $R_{14n} \equiv \frac{1}{n} \sum_{i=1}^n \sum_{j=1}^n \{r_i \circ [\hat{r}(Y_j; X_i, Z_j) - r(Y_j; X_i, Z_j)] \circ r_{2i}\} a_i$  and  $R_{15n} \equiv \frac{1}{n} \sum_{i=1}^n \sum_{j=1}^n \{r_i \circ [r(Y_j; X_i, Z_j) - r_1(X_i)] \circ r_{2i}\} a_i$ . By straightforward moment calculations, we can show that  $R_{15n} = o_P(n^{1/2})$ . By Lemma B.3(a), we can show that  $R_{14n} = \bar{R}_{14n} + o_P(n^{1/2}b^{-(\frac{d}{2}+1)})$ , where

$$\bar{R}_{14n} = \frac{1}{n^2} \sum_{i=1}^n \sum_{j=1}^n \sum_{k=1}^n [r_i \circ \zeta_k(Y_j; X_i, Z_j) \circ r_{2i}] a_i.$$

Noting that  $E \|\bar{R}_{14n}\|^2 = O(nb^{-2} + (b^{-d_x-2} + b^{-2d_z-2} + c^{-d_x-1} + c^{-2d_z-2}) + n^{-1}(b^{-d-2-d_z} + c^{-d-2-d_z}))$ , we have  $\bar{R}_{14n} = o_P(n^{1/2}b^{-1} + b^{-d_z-1} + c^{-d_z-1})$ . Consequently,  $\bar{\Gamma}_{2n,8} = o_P(n^{1/2}b^{-1} + b^{-d_z-1} + c^{-d_z-1}) + o_P(n^{1/2}b^{-(\frac{d}{2}+1)})$ . Then by Lemma B.3(b) and the fact that  $\nu_{bc} = o_P(n^{-1/2}b^{-1})$ ,

$$\begin{aligned} \Gamma_{2n,8} &= b^{\frac{d}{2}+2} o_P(n^{-1/2}b^{-1}) \left[ o_P(n^{1/2}b^{-1} + b^{-d_z-1} + c^{-d_z-1}) + o_P(n^{1/2}b^{-(\frac{d}{2}+1)}) \right] \\ &= o_P(b^{\frac{d}{2}} + n^{-1/2}b^{\frac{d_x-d_z}{2}} + n^{-1/2}b^{\frac{d}{2}+2}c^{-d_z-1}) + o_P(1) = o_P(1). \end{aligned}$$

(e) By the Cauchy-Schwarz inequality and Lemmas B.5(a) and (c),  $|\Gamma_{2n,9}| \leq 2(\Gamma_{2n,2}\Gamma_{2n,4})^{1/2} = o_P(1)$ .

(f) Let  $\bar{\Delta}r_{1ij} = \hat{r}(Y_j; X_i, Z_j) - r(Y_j; X_i, Z_j)$ ,  $\Delta r_{1ij} = r(Y_j; X_i, Z_j) - r_1(X_i)$ ,  $\bar{\Delta}r_{2ij} = \hat{r}(Y_i; X_j, Z_j) - r(Y_i; X_j, Z_j)$ , and  $\Delta r_{2ij} = r(Y_i; X_j, Z_j) - r_2(Y_i)$ . Using  $\hat{r}_{1i} - r_{1i} = \frac{1}{n} \sum_{j=1}^n (\bar{\Delta}r_{1ij} + \Delta r_{1ij})$  and  $\hat{r}_{2i} - r_{2i} = \frac{1}{n} \sum_{k=1}^n (\bar{\Delta}r_{2ik} + \Delta r_{2ik})$ , we can decompose  $\Gamma_{2n,10}$  as follows:

$$\begin{aligned} \Gamma_{2n,10} &= 2n^{-2}b^{\frac{d}{2}+2} \sum_{i=1}^n \sum_{j=1}^n \sum_{k=1}^n \left\{ (\Delta r_{1ij} \circ r_{2i})' (r_{1i} \circ \Delta r_{2ik}) a_i + (\bar{\Delta}r_{1ij} \circ r_{2i})' (r_{1i} \circ \bar{\Delta}r_{2ik}) a_i \right. \\ &\quad \left. + (\bar{\Delta}r_{1ij} \circ r_{2i})' (r_{1i} \circ \Delta r_{2ik}) a_i + (\Delta r_{1ij} \circ r_{2i})' (r_{1i} \circ \bar{\Delta}r_{2ik}) a_i \right\} \\ &\equiv 2R_{16n} + 2R_{17n} + 2R_{18n} + 2R_{19n}, \text{ say.} \end{aligned}$$

By moment calculations,  $E(R_{16n}) = O(b^{\frac{d}{2}+2})$  and  $E(R_{16n}^2) = O(b^{d+4})$ , implying that  $R_{16n} = o_P(b^{\frac{d}{2}+2}) = o_P(1)$ . For  $R_{17n}$ , we can show that  $R_{17n} = \bar{R}_{17n} + o_P(1)$ , where

$$\bar{R}_{17n} = n^{-4}b^{\frac{d}{2}+2} \sum_{i_1=1}^n \sum_{i_2=1}^n \sum_{i_3=1}^n \sum_{i_4=1}^n \sum_{i_5=1}^n (\zeta_{i_4}(Y_{i_2}; X_{i_1}, Z_{i_2}) \circ r_{2i_1})' (r_{1i_1} \circ \zeta_{i_5}(Y_{i_1}; X_{i_3}, Z_{i_3})) a_{i_1}.$$

Noting that  $E(\bar{R}_{17n}^2) = O(b^{d+4}b^{-4}) + n^{-1}b^{d+4}(b^{-d_x-4} + b^{-d_z-4} + c^{-d_x} + c^{-d_z}) = o(1)$ , we have  $\bar{R}_{17n} = o_P(1)$ . Similarly, we can show that  $R_{18n} = o_P(1)$  and  $R_{19n} = o_P(1)$ . Consequently,  $\Gamma_{2n,10} = o_P(1)$ . ■

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