

Functional Linear Regression Model for Nonignorable Missing Scalar Responses

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Supplementary Material

S1 Proofs

Lemma 1–Lemma 9 and Lemma 10–Lemma 13 are listed in the following for proofs of Theorem 1 and 2, respectively. Proofs of Lemma 1, 2, and Lemma 3- (i) can be found in Lemma 3.3 of Hall and Hosseini-Nasab (2009), Theorem 3 of Hall and Hosseini-Nasab (2006), and Proposition 18 of Crambes and Andr (2013), respectively. Before continuing, we define the following operators and notations.

First we define $x_n \preceq y_n$ or $x_n = O_p(y_n)$ for random sequences (x_n) and (y_n) , if for any $\tau > 0$, there exist $M_\tau > 0$, and $N > 0$ such that for any $n > N$, $\Pr(|x_n/y_n| > M_\tau) < \tau$; $x_n \succeq y_n$, or $y_n = O_p(x_n)$, if for any $\tau > 0$, there exists $M_\tau > 0$ such that $\Pr(|y_n/x_n| > M_\tau) < \tau$; $x_n \ll y_n$ or $x_n = o_p(y_n)$, if for any $\tau > 0$, $\Pr(|x_n/y_n| > \tau) \rightarrow 0$; $x_n \gg y_n$ or $y_n = o_p(x_n)$, if for any $\tau > 0$, $\Pr(|y_n/x_n| > \tau) \rightarrow 0$; $x_n \sim y_n$, if $x_n \preceq y_n$ as well as $x_n \succeq y_n$. Second, for an arbitrary bivariate function $f : \mathfrak{D}_x \times \mathfrak{D}_y \mapsto \mathfrak{R}$, random variables $\xi : \Omega \mapsto \mathfrak{D}_x$, and $\eta : \Omega \mapsto \mathfrak{D}_y$, if $E[f(x, \eta)] < \infty$ for any $x \in \mathfrak{D}_x$, define the notation $E_{-\xi}f(\xi, \eta)$ by $E_{-\xi}f(\xi, \eta) = g(\xi)$ where $g : \mathfrak{D}_x \mapsto \mathfrak{R}$, $g(x) = E[f(x, \eta)]$ for any $x \in \mathfrak{D}_x$. Note that if ξ and η are independent, $E_{-\xi}f(\xi, \eta) = E[f(\xi, \eta)|\xi]$. Third, we denote

$$r_j^* = \frac{\sum_{i=1}^n [\delta_i M_j(Y_i, Z_i, W_i, v_j; \beta_{1,0}) + (1 - \delta_i) m_{M_j, i, \gamma}^0(Y_i, Z_i, W_i, v_j; \beta_{1,0})]}{n\lambda_j},$$

$$\hat{r}_j^* = \frac{\sum_{i=1}^n [\delta_i M_j(Y_i, Z_i, W_i, v_j; \beta_{1,0}) + (1 - \delta_i) \hat{m}_{M_j, i, \gamma}(Y_i, Z_i, W_i, v_j; \beta_{1,0})]}{n\lambda_j},$$

$\gamma_0 = -\phi_0$, $\zeta_j = \min_{k \leq j} |\lambda_k - \lambda_{k+1}|$, and define $F(\langle Z_l, \theta_0 \rangle, G(\langle Z_l, W_l \rangle), W_l)$ as the following conditional expectation

$$E(E[\langle Z_l, \theta_0 \rangle \delta_l \exp(\gamma_0 Y_l) | Z_l, W_l] | \langle Z_l, \theta_0 \rangle, G(\langle Z_l, W_l \rangle), W_l).$$

Furthermore, since $Z_i, i = 1, 2, \dots, n$ are n independent and identically distributed realizations of Z , from Assumption (A.2), there exist random variables $\xi_j^{(i)}, i \leq n, j \in \mathbb{Z}_+$ such that $Z_i = \sum_{j=1}^{\infty} \sqrt{\lambda_j} \xi_j^{(i)} v_j, i, j \in \mathbb{Z}_+$, where $\xi_j^{(i)}, i = 1, 2, \dots, n$ are n independent and identically distributed realizations of ξ_j , and $\xi_j^{(i)}$ are mutually independent for $i \leq n, j \in \mathbb{Z}_+$. Finally, for a kernel function $K : R \mapsto [0, +\infty)$, and $K_h : R \mapsto [0, +\infty), K_h(\cdot) = K(\cdot/h)$ we define random functions $K_h^{(l)}(\cdot)$ as

$$K_h^{(l)}(\cdot) = K_h(\cdot) \delta_l \exp(\gamma_0 Y_l) \langle Z_l, \boldsymbol{\theta}_0 \rangle,$$

and $\tilde{K}_h^{(l)}(\cdot)$ as

$$\tilde{K}_h^{(l)}(\cdot) = K_h(\cdot) \delta_l \exp(\gamma_0 Y_l),$$

for $l = 1, 2, \dots, n$ and $j \in \mathbb{Z}_+$.

Lemma 1. *Assume that with probability 1, X is left-continuous at each point (or right-continuous at each point), and that Conditions (B3) and (B4) hold. Then, for each $C > 0$, $E(\|\hat{\mathcal{K}} - \mathcal{K}\|^C) < \text{constant} * n^{-C/2}$, where \mathcal{K} and $\hat{\mathcal{K}}$ are the covariance and the sample covariance function of the process $Z(\cdot)$, and*

$$\|\hat{\mathcal{K}} - \mathcal{K}\| \triangleq \sqrt{\int_{[0,1]^2} [\hat{\mathcal{K}}(s_1, s_2) - \mathcal{K}(s_1, s_2)]^2 ds_1 ds_2}$$

Lemma 2 *Under Assumptions (A.2) and (A.3), we have*

$$\|\hat{v}_j - v_j\| \leq 8^{1/2} \zeta_j^{-1} \|\hat{\mathcal{K}} - \mathcal{K}\| \text{ for any } j.$$

Lemma 3

(i) *Under Assumptions (A.2) and (A.3), when j is large enough,*

$$0 < \text{constant} \times j^{-a} \leq \lambda_j < \text{constant} \times j^{-1};$$

(ii) *under Assumptions (A.2) and (A.5), we have*

$$\Pr(\overline{\lim}_{k \rightarrow \infty} \bigcup_{j \leq k} |\hat{\lambda}_j - \lambda_j| > (\lambda_j - \lambda_{j+1})/2) = 0,$$

which implies

$$\Pr(\overline{\lim}_{k \rightarrow \infty} \bigcup_{j \leq k} \hat{\lambda}_j < \lambda_{j+1}) = 0.$$

Proof of Lemma 3, (i).

From $E\langle Z, Z \rangle = \sum_{j=1}^{\infty} \langle Z, v_j \rangle^2 = \sum_{j=1}^{\infty} \lambda_j < \infty$, we have $\lambda_j \ll j^{-1}$; from $\lambda_j = \sum_{k=j}^{\infty} (\lambda_k - \lambda_{k+1})$ and Assumption (A.3) we have $\lambda_j \gg j^{-a}$.

□

Lemma 4. *Under Assumptions (A.1), (A.2), (A.6) and (A.9), we have*

(i)

$$m_{M, i, \gamma}^0(Y_i, W_i; \boldsymbol{\beta}_{1,0}) = E[\widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) | \delta_i = 0, Z_i, W_i];$$

(ii)

$$L_4 \triangleq \frac{1}{n} \sum_{i=1}^n [\delta_i \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) + (1 - \delta_i) m_{\widetilde{M}, i, \gamma}^0(Y_i, W_i; \boldsymbol{\beta}_{1,0})] - E[\langle \boldsymbol{\theta}_0, Z \rangle W] = O_p(1/\sqrt{n}).$$

Proof. Part (i) is shown in the following.

$$\begin{aligned} m_{\widetilde{M}, i, \gamma}^0(Y_i, W_i; \boldsymbol{\beta}_{1,0}) &= \frac{E\{\delta_i \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) \exp(\gamma Y_i) | Z_i, W_i\}}{E\{\delta_i \exp(\gamma Y_i) | Z_i, W_i\}} \\ &= \frac{E\{E[\delta_i \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) \exp(\gamma Y_i) | Z_i, Y_i, W_i] | Z_i, W_i\}}{E\{E[\delta_i \exp(\gamma Y_i) | Z_i, Y_i, W_i] | Z_i, W_i\}} \\ &= \frac{E\{E[\delta_i | Z_i, W_i, Y_i] \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) \exp(\gamma Y_i) | Z_i, W_i\}}{E\{E[\delta_i | Z_i, W_i, Y_i] \exp(\gamma Y_i) | Z_i, W_i\}} \\ &= \frac{E\{\Pr(\delta_i = 1 | Y_i, Z_i, W_i) \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) \exp(\gamma Y_i) | Z_i, W_i\}}{E\{\Pr(\delta_i = 1 | Y_i, Z_i, W_i) \exp(\gamma Y_i) | Z_i, W_i\}} \\ &= \frac{E\{\Pr(\delta_i = 0 | Y_i, Z_i, W_i) \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) | Z_i, W_i\}}{E\{\Pr(\delta_i = 0 | Y_i, Z_i, W_i) | Z_i, W_i\}} \\ &= \frac{E\{(1 - \delta_i) \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) | Z_i, W_i\}}{E\{(1 - \delta_i) | Z_i, W_i\}} = E[\widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) | \delta_i = 0, Z_i, W_i]. \end{aligned}$$

To prove (ii), first we calculate the expectation as below.

$$\begin{aligned} &E[\delta_i \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) + (1 - \delta_i) m_{\widetilde{M}, i, \gamma}^0(Y_i, W_i; \boldsymbol{\beta}_{1,0})] \\ &= E\left\{\delta_i \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) + (1 - \delta_i) E[\widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) | \delta_i = 0, Z_i, W_i]\right\} \\ &= \Pr(\delta_i = 1) E[\widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) | \delta_i = 1] + \Pr(\delta_i = 0) E\{E[\widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) | \delta_i = 0, Z_i, W_i] | \delta_i = 0\} \\ &= \Pr(\delta_i = 1) E[\widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) | \delta_i = 1] + \Pr(\delta_i = 0) E\{\widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) | \delta_i = 0\} \\ &= E[\widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0})] = E[\langle \boldsymbol{\theta}_0, Z \rangle W]. \end{aligned}$$

Second we calculate the variance, using the independence across different subjects.

$$\begin{aligned} &E^2\left\{\frac{1}{n} \sum_{i=1}^n [\delta_i \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) + (1 - \delta_i) m_{\widetilde{M}, i, \gamma}^0(Y_i, W_i; \boldsymbol{\beta}_{1,0})] - E[\langle \boldsymbol{\theta}_0, Z \rangle W]\right\} \\ &= \frac{1}{n^2} E^2 \sum_{i=1}^n \{[\delta_i \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) + (1 - \delta_i) m_{\widetilde{M}, i, \gamma}^0(Y_i, W_i; \boldsymbol{\beta}_{1,0})] - E[\langle \boldsymbol{\theta}_0, Z \rangle W]\} \\ &= \frac{1}{n} E^2\{[\delta_i \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) + (1 - \delta_i) m_{\widetilde{M}, i, \gamma}^0(Y_i, W_i; \boldsymbol{\beta}_{1,0})] - E[\langle \boldsymbol{\theta}_0, Z \rangle W]\} \\ &\leq \frac{1}{n} E^2\{\delta_i \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) + (1 - \delta_i) m_{\widetilde{M}, i, \gamma}^0(Y_i, W_i; \boldsymbol{\beta}_{1,0})\} \\ &= \frac{1}{n} E^2\{\delta_i \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) + (1 - \delta_i) E[\widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) | \delta_i = 0, Z_i, W_i]\} \\ &\leq \frac{2}{n} E^2[\widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0})] = O_p(1/n). \end{aligned}$$

Finally, we have

$$\frac{1}{n} \sum_{i=1}^n [\delta_i \widetilde{M}(Y_i, W_i; \boldsymbol{\beta}_{1,0}) + (1 - \delta_i) m_{\widetilde{M}, i, \gamma}^0(Y_i, W_i; \boldsymbol{\beta}_{1,0})] = E[\langle \boldsymbol{\theta}_0, Z \rangle W] + O_p(1/\sqrt{n}).$$

□

Lemma 5. *Suppose (Z, W, Y, δ) is independently and identically distributed with $(Z_i, W_i, Y_i, \delta_i)$, $i = 1, 2, \dots, n$. Then under Assumptions (A.1), (A.2), (A.4), (A.6), (A.7), and (A.9), we have*

$$(i) \quad \mathbb{E}r_j^* = \langle \boldsymbol{\theta}_0, v_j \rangle.$$

$$(ii) \quad L_5 \triangleq \mathbb{E} \sum_{j=1}^{k_n} r_j^* [\delta W \langle Z, v_j \rangle + (1 - \delta) \frac{\mathbb{E}\{\delta \langle Z, v_j \rangle W \exp(\gamma Y) | X, V\}}{\mathbb{E}\{\delta \exp(\gamma Y) | X, V\}}] - \mathbb{E}[\langle \boldsymbol{\theta}_0, Z \rangle W] = O(k_n^{1/2-b}).$$

Proof. Similar to the proof of Lemma 1, we have

$$\mathbb{E}r_j^* = \mathbb{E}M_j(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1) / \lambda_j = \langle \boldsymbol{\theta}_0, v_j \rangle; \quad (\text{S1.1})$$

$$\begin{aligned} U_1 &\triangleq \mathbb{E} \sum_{j=1}^{k_n} \langle \boldsymbol{\theta}_0, v_j \rangle [\delta W \langle Z, v_j \rangle + (1 - \delta) \frac{\mathbb{E}\{\delta \langle Z, v_j \rangle W \exp(\gamma Y) | X, V\}}{\mathbb{E}\{\delta \exp(\gamma Y) | X, V\}}] \\ &= \sum_{j=1}^{k_n} \langle \boldsymbol{\theta}_0, v_j \rangle \mathbb{E}[\langle Z, v_j \rangle W] = \mathbb{E}[\langle \boldsymbol{\theta}_0, Z \rangle W] - \sum_{j=k_n+1}^{\infty} \langle \boldsymbol{\theta}_0, v_j \rangle \mathbb{E}[\langle Z, v_j \rangle W]. \end{aligned} \quad (\text{S1.2})$$

Hence (i) has been proved. To prove (ii), using the independence of (Z, W, Y, δ) with $(Z_i, W_i, Y_i, \delta_i)$, $i = 1, 2, \dots, n$, and following (??), we have

$$\begin{aligned} L_5 &= \mathbb{E} \sum_{j=1}^{k_n} \mathbb{E}(r_j^*) [\delta W \langle Z, v_j \rangle + (1 - \delta) \frac{\mathbb{E}\{\delta \langle Z, v_j \rangle W \exp(\gamma Y) | X, V\}}{\mathbb{E}\{\delta \exp(\gamma Y) | X, V\}}] - \mathbb{E}[\langle \boldsymbol{\theta}_0, Z \rangle W] \\ &= U_1 - \mathbb{E}[\langle \boldsymbol{\theta}_0, Z \rangle W] = - \sum_{j=k_n+1}^{\infty} \langle \boldsymbol{\theta}_0, v_j \rangle \mathbb{E}[\langle Z, v_j \rangle W]. \end{aligned}$$

It follows that

$$\begin{aligned} |L_5| &= \sum_{j=k_n+1}^{\infty} \langle \boldsymbol{\theta}_0, v_j \rangle \mathbb{E}[\langle Z, v_j \rangle W] = \sum_{j=k_n+1}^{\infty} \sqrt{\lambda_j} \langle \boldsymbol{\theta}_0, v_j \rangle \mathbb{E}(\xi_j W) \\ &\leq \sum_{j=k_n+1}^{\infty} \sqrt{\lambda_j} \langle \boldsymbol{\theta}_0, v_j \rangle \sqrt{\mathbb{E}(W^2) \mathbb{E}\xi_j^2} = \sqrt{\mathbb{E}W^2} \sum_{j=k_n+1}^{\infty} \sqrt{\lambda_j} \langle \boldsymbol{\theta}_0, v_j \rangle. \end{aligned}$$

Together with $\lambda_j \preceq j^{-1}$ in Lemma 3 and $\langle \boldsymbol{\theta}_0, v_j \rangle \preceq j^{-b}$ in Assumption (A.4), we have $L_5 = O(k_n^{1/2-b})$. □

Lemma 6. *Under Assumptions (A.1)–(A.7), and (A.9), we have*

$$(i) \quad \sum_{j=1}^{k_n} j^x \lambda_j (r_j^* - \langle \boldsymbol{\theta}_0, v_j \rangle)^2 = O_p(k_n^{1+x}/n), \text{ for any } x > -1.$$

$$(ii) \quad \sup_{j \leq k_n} \zeta_j \lambda_j |\Delta_2(r_j)| = O_p(k_n^{a-1}/\sqrt{n}),$$

where

$$\Delta_2(r_j) = r_j - \frac{\sum_{i=1}^n [\delta_i M_j(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1) + (1 - \delta_i) m_{M_j, i, \gamma}^0(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1)]}{n \hat{\lambda}_j}.$$

(iii)

$$\sum_{j=1}^{k_n} \lambda_j |r_j - r_j^*|^2 = O_p(k_n^{4a+2} n^{-1} + k_n^{-2b});$$

(iv)

$$\sum_{j=1}^{k_n} \mathbb{E} \left[\frac{1}{n} \sum_{i=1}^n (\delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) \mathbb{E}(W_i \langle Z_i, v_j \rangle | Z_i, W_i, \delta_i = 0)) \right]^2 = O(1);$$

(v)

$$\sup_j \zeta_j \left| \frac{1}{n} \sum_{i=1}^n \delta_i W_i \langle Z_i, \hat{v}_j - v_j \rangle \right| + \sup_j \frac{1}{n} \sum_{i=1}^n \mathbb{E}_{-\hat{v}_j} [\zeta_j |W_i \langle Z_i, \hat{v}_j - v_j \rangle| | Z_i, W_i, \delta_i = 0] = O_p(1/\sqrt{n}).$$

(vi) $L_6 \triangleq$

$$\frac{1}{n} \sum_{i=1}^{k_n} \Delta_1 \left\{ \sum_{j=1}^{k_n} r_j [\delta_i W_i \langle Z_i, \hat{v}_j \rangle + (1 - \delta_i) \mathbb{E}_{-\hat{v}_j} \{ \langle Z_i, \hat{v}_j \rangle W | Z_i, W_i, \delta_i = 0 \}] \right\} = O_p(k_n^{2a+1} n^{-1/2} + k_n^{-b}),$$

where

$$\begin{aligned} & \Delta_1 \left\{ \sum_{j=1}^{k_n} r_j [\delta_i W_i \langle Z_i, \hat{v}_j \rangle + (1 - \delta_i) \mathbb{E}_{-\hat{v}_j} \{ \langle Z_i, \hat{v}_j \rangle W | Z_i, W_i, \delta_i = 0 \}] \right\} \\ &= \left\{ \sum_{j=1}^{k_n} r_j [\delta_i W_i \langle Z_i, \hat{v}_j \rangle + (1 - \delta_i) \mathbb{E}_{-\hat{v}_j} \{ \langle Z_i, \hat{v}_j \rangle W | Z_i, W_i, \delta_i = 0 \}] \right\} \\ &- \left\{ \sum_{j=1}^{k_n} r_j^* [\delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) \mathbb{E} \{ \langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0 \}] \right\}. \end{aligned}$$

Proof of Lemma 6 (i). Using conclusion 1 of Lemma 5, we have

$$\begin{aligned} & \mathbb{E} \sum_{j=1}^{k_n} j^x \lambda_j (r_j^* - \langle \boldsymbol{\theta}_0, v_j \rangle)^2 \\ &= \mathbb{E} \sum_{j=1}^{k_n} j^x \left(\sum_{i=1}^n \frac{\delta_i M_j(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1) + (1 - \delta_i) m_{M_j, i, \gamma}^0(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1)}{n \sqrt{\lambda_j}} - \sqrt{\lambda_j} \langle \boldsymbol{\theta}_0, v_j \rangle \right)^2 \\ &= \sum_{j=1}^{k_n} j^x \frac{1}{n^2} \mathbb{E} \sum_{i=1}^n \left(\frac{\delta_i (M_j(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1) - \lambda_j \langle \boldsymbol{\theta}_0, v_j \rangle)}{\sqrt{\lambda_j}} \right. \\ &+ \left. \frac{\sum_{i=1}^n (1 - \delta_i) (m_{M_j, i, \gamma}^0(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1) - \lambda_j \langle \boldsymbol{\theta}_0, v_j \rangle)}{\sqrt{\lambda_j}} \right)^2 \\ &\leq \frac{2}{n} \sum_{j=1}^{k_n} j^x \mathbb{E} \left(\frac{\delta_i (M_j(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1) - \lambda_j \langle \boldsymbol{\theta}_0, v_j \rangle)}{\sqrt{\lambda_j}} \right)^2 \\ &+ \frac{2}{n} \sum_{j=1}^{k_n} j^x \mathbb{E} \left(\frac{(1 - \delta_i) (m_{M_j, i, \gamma}^0(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1) - \lambda_j \langle \boldsymbol{\theta}_0, v_j \rangle)}{\sqrt{\lambda_j}} \right)^2 \triangleq \frac{2}{n} (A^{(6,1)} + B^{(6,1)}). \end{aligned}$$

In the following the order of $A^{(6,1)}$ and $B^{(6,1)}$ are calculated separately.

$$\begin{aligned}
EA^{(6,1)} &= \sum_{j=1}^{k_n} j^x \mathbb{E} \left(\frac{\delta_i(M_j(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1) - \lambda_j \langle \boldsymbol{\theta}_0, v_j \rangle)}{\sqrt{\lambda_j}} \right)^2 \\
&\leq \sum_{j=1}^{k_n} j^x 2 \mathbb{E} \left[\left(\frac{M_j(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1)}{\sqrt{\lambda_j}} \right)^2 + \left(\frac{\lambda_j \langle \boldsymbol{\theta}_0, v_j \rangle}{\sqrt{\lambda_j}} \right)^2 \right] \\
&\leq \sum_{j=1}^{k_n} j^x 2 \left[\mathbb{E} \left(\frac{\langle Z, v_j \rangle (\langle \boldsymbol{\theta}, Z \rangle + \epsilon)}{\sqrt{\lambda_j}} \right)^2 + (j^{-1/2-b})^2 \right] \\
&= \sum_{j=1}^{k_n} j^x 2 \left[\mathbb{E} (\xi_j (\langle \boldsymbol{\theta}, Z \rangle + \epsilon))^2 + (j^{-1/2-b})^2 \right] \\
&\leq \left(\sqrt{\mathbb{E} \xi_1^4} \sqrt{\mathbb{E} (\langle \boldsymbol{\theta}, Z \rangle + \epsilon)^4} + \text{constant} \right) \sum_{j=1}^{k_n} j^x = O(k_n^{1+x}); \\
EB^{(6,1)} &= \sum_{j=1}^{k_n} j^x \mathbb{E} \left(\frac{(1 - \delta_i) (\mathbb{E}(M_j(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1) | Z_i, W_i, \delta_i = 0) - \lambda_j \langle \boldsymbol{\theta}_0, v_j \rangle)}{\sqrt{\lambda_j}} \right)^2 \\
&\leq \sum_{j=1}^{k_n} j^x 2 \mathbb{E} \left[\left(\frac{\mathbb{E}(M_j(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1) | Z_i, W_i, \delta_i = 0)}{\sqrt{\lambda_j}} \right)^2 + \left(\frac{\lambda_j \langle \boldsymbol{\theta}_0, v_j \rangle}{\sqrt{\lambda_j}} \right)^2 \right] \\
&\leq \sum_{j=1}^{k_n} j^x 2 \mathbb{E} \left[\mathbb{E} \left(\frac{M_j^2(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1)}{\lambda_j} | Z_i, W_i, \delta_i = 0 \right) + \left(\frac{\lambda_j \langle \boldsymbol{\theta}_0, v_j \rangle}{\sqrt{\lambda_j}} \right)^2 \right] \\
&\leq \sum_{j=1}^{k_n} j^x 2 \mathbb{E} \left[\left(\frac{M_j(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1)}{\sqrt{\lambda_j}} \right)^2 + \left(\frac{\lambda_j \langle \boldsymbol{\theta}_0, v_j \rangle}{\sqrt{\lambda_j}} \right)^2 \right] = O(k_n^{1+x}),
\end{aligned}$$

Combine them together, and we have $\mathbb{E} \sum_{j=1}^{k_n} j^x \lambda_j (r_j^* - \langle \boldsymbol{\theta}_0, v_j \rangle)^2 = O((A^{(6,1)} + B^{(6,1)})/n) = O(k_n^{1+x}/n)$. \square

Proof of Lemma 6 (ii). Denote

$$\Delta_2(M_j(Y_i, Z_i, W_i, \hat{v}_j; \boldsymbol{\beta}_1)) = M_j(Y_i, Z_i, W_i, \hat{v}_j; \boldsymbol{\beta}_1) - M_j(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1),$$

and

$$\Delta_2(m_{M_j, i, \gamma}^0(Y_i, Z_i, W_i, \hat{v}_j; \boldsymbol{\beta}_1)) = m_{M_j, i, \gamma}^0(Y_i, Z_i, W_i, \hat{v}_j; \boldsymbol{\beta}_1) - m_{M_j, i, \gamma}^0(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1).$$

From Lemma 3 (i) we have $\lambda_j/\lambda_{j+1} \leq k_n^{a-1}$ when j is sufficiently large. Then $\sup_{j \leq k_n} \lambda_j/\lambda_{j+1} \leq k_n^{a-1}$ when k_n is sufficiently large. Together with Lemma 3 (ii) we have $\mathbb{E} \sup_{j \leq k_n} \zeta_j \lambda_j |\Delta_2(r_j)| \leq$

$$\begin{aligned}
&\mathbb{E} \sup_{j \leq k_n} \frac{1}{n} \sum_{i=1}^n \zeta_j \frac{\lambda_j}{\lambda_j} [|\delta_i \Delta_2(M_j(Y_i, Z_i, W_i, \hat{v}_j; \boldsymbol{\beta}_1))| + |(1 - \delta_i) \Delta_2(m_{M_j, i, \gamma}^0(Y_i, Z_i, W_i, \hat{v}_j; \boldsymbol{\beta}_1))|] \\
&\leq \frac{1}{n} \sum_{i=1}^n \mathbb{E} \sup_{j \leq k_n} \zeta_j \frac{\lambda_j}{\lambda_{j+1}} [|\delta_i \Delta_2(M_j(Y_i, Z_i, W_i, \hat{v}_j; \boldsymbol{\beta}_1))| + |(1 - \delta_i) \Delta_2(m_{M_j, i, \gamma}^0(Y_i, Z_i, W_i, \hat{v}_j; \boldsymbol{\beta}_1))|] \\
&= \mathbb{E} \sup_{j \leq k_n} \zeta_j \frac{\lambda_j}{\lambda_{j+1}} [|\delta_i \Delta_2(M_j(Y_i, Z_i, W_i, \hat{v}_j; \boldsymbol{\beta}_1))| + |(1 - \delta_i) \Delta_2(m_{M_j, i, \gamma}^0(Y_i, Z_i, W_i, \hat{v}_j; \boldsymbol{\beta}_1))|] \\
&\leq \mathbb{E} \sup_{j \leq k_n} \zeta_j \frac{\lambda_j}{\lambda_{j+1}} |\Delta_2(M_j(Y_i, Z_i, W_i, \hat{v}_j; \boldsymbol{\beta}_1))| + \mathbb{E} \sup_{j \leq k_n} \zeta_j \frac{\lambda_j}{\lambda_{j+1}} |\Delta_2(m_{M_j, i, \gamma}^0(Y_i, Z_i, W_i, \hat{v}_j; \boldsymbol{\beta}_1))|
\end{aligned}$$

$$\begin{aligned}
&\leq \sup_{j \leq k_n} \frac{\lambda_j}{\lambda_{j+1}} [\mathbb{E} \sup_{j \leq k_n} \zeta_j |\Delta_2(M_j(Y_i, Z_i, W_i, \hat{v}_j; \beta_1))| + \mathbb{E} \sup_{j \leq k_n} \zeta_j |\Delta_2(m_{M_j, i, \gamma}^0(Y_i, Z_i, W_i, \hat{v}_j; \beta_1))|] \\
&\leq \text{constant} * k_n^{a-1} [\mathbb{E} \sup_{j \leq k_n} \zeta_j |\Delta_2(M_j(Y_i, Z_i, W_i, \hat{v}_j; \beta_1))| + \mathbb{E} \sup_{j \leq k_n} \zeta_j |\Delta_2(m_{M_j, i, \gamma}^0(Y_i, Z_i, W_i, \hat{v}_j; \beta_1))|] \\
&\triangleq \text{constant} * k_n^{a-1} (A^{(6,2)} + B^{(6,2)}).
\end{aligned}$$

Next the two terms $A^{(6,2)}$ and $B^{(6,2)}$ are calculated separately. It follows from the Cauchy's inequality that

$$\begin{aligned}
A^{(6,2)} &= \mathbb{E} \sup_{j \leq k_n} \zeta_j |\langle Z_i, \hat{v}_j - v_j \rangle \langle Z_i, \boldsymbol{\theta} \rangle + \epsilon_i| \\
&= \mathbb{E} \sup_{j \leq k_n} |(\langle Z_i, \boldsymbol{\theta} \rangle + \epsilon_i) Z_i, \zeta_j (\hat{v}_j - v_j)| \\
&\leq \mathbb{E} \sqrt{\sup_{j \leq k_n} |(\langle Z_i, \boldsymbol{\theta} \rangle + \epsilon_i) Z_i, (\langle Z_i, \boldsymbol{\theta} \rangle + \epsilon_i) Z_i| \langle \zeta_j (\hat{v}_j - v_j), \zeta_j (\hat{v}_j - v_j) \rangle} \\
&= \sqrt{\mathbb{E} \langle Z, Z \rangle \langle \langle Z, \boldsymbol{\theta} \rangle + \epsilon \rangle^2 \{ \mathbb{E} \sup_j \zeta_j^2 \|\hat{v}_j - v_j\|^2 \}^{1/2}} (1 + o(1)).
\end{aligned}$$

From Lemma 1 and Lemma 2, $\{ \mathbb{E} \sup_{j \leq k_n} \zeta_j^2 \|\hat{v}_j - v_j\|^2 \}^{1/2} \leq \text{constant} * n^{-1/2}$. Then $A^{(6,2)} = O_p(1/\sqrt{n})$. Using Lemma 4, by Jensen's inequality, we have

$$\begin{aligned}
B^{(6,2)} &= \mathbb{E} \left\{ \sup_{j \leq k_n} \zeta_j |\mathbb{E}_{-\hat{v}_j} [\langle Z_i, \hat{v}_j - v_j \rangle \langle Z_i, \boldsymbol{\theta} \rangle + \epsilon_i] | Z_i, W_i, \delta_i = 0] \right\} \\
&\leq \mathbb{E} \left\{ \mathbb{E}_{-(\hat{v}_1, \dots)} \left[\sup_{j \leq k_n} \zeta_j |\langle Z_i, \hat{v}_j - v_j \rangle \langle Z_i, \boldsymbol{\theta} \rangle + \epsilon_i| | Z_i, W_i, \delta_i = 0] \right\} \\
&= \mathbb{E} \left\{ \mathbb{E}_{-(\hat{v}_1, \dots)} \left[\sup_{j \leq k_n} |(\langle Z_i, \boldsymbol{\theta} \rangle + \epsilon_i) Z_i, \zeta_j (\hat{v}_j - v_j)| | Z_i, W_i, \delta_i = 0] \right\} \\
&\leq \mathbb{E} \sqrt{\langle Z, Z \rangle \langle \langle Z, \boldsymbol{\theta} \rangle + \epsilon \rangle^2 \{ \mathbb{E} \mathbb{E}_{-(\hat{v}_1, \dots)} \sup_j \zeta_j^2 \|\hat{v}_j - v_j\|^2 \}^{1/2}} (1 + o(1)) \\
&= \mathbb{E} \sqrt{\langle Z, Z \rangle \langle \langle Z, \boldsymbol{\theta} \rangle + \epsilon \rangle^2 \{ \mathbb{E} \sup_j \zeta_j^2 \|\hat{v}_j - v_j\|^2 \}^{1/2}} (1 + o(1)).
\end{aligned}$$

Then $B^{(6,2)} = O_p(1/\sqrt{n})$. □

Proof of Lemma 6 (iii). From definitions of r_j , r_j^* , and $\Delta_2(r_j)$ in Lemma 6, (ii), we have

$$|r_j - r_j^*| \leq |\Delta_2(r_j)| + \left| \frac{1}{\lambda_j} - \frac{1}{\hat{\lambda}_j} \right| |\hat{\lambda}_j r_j^*|.$$

It follows that

$$\begin{aligned}
&\left(\sum_{j=1}^{k_n} \lambda_j |r_j - r_j^*|^2 \right) \leq \left[\sum_{j=1}^{k_n} \lambda_j 2(\Delta_2(r_j))^2 + \left(\frac{1}{\lambda_j} - \frac{1}{\hat{\lambda}_j} \right)^2 \hat{\lambda}_j^2 r_j^{*2} \right] \\
&= \sum_{j=1}^{k_n} 2\lambda_j \Delta_2(r_j)^2 + \sum_{j=1}^{k_n} 2 \left(\frac{1}{\lambda_j} - \frac{1}{\hat{\lambda}_j} \right)^2 \hat{\lambda}_j^2 r_j^{*2} \lambda_j \triangleq A^{(6,3)} + B^{(6,3)}.
\end{aligned}$$

Next the two terms $A^{(6,3)}$ and $B^{(6,3)}$ are calculated. From Lemma 3, we have $EA^{(6,3)} \leq$

$$\begin{aligned}
& 2E\left(\sum_{j=1}^{k_n} \lambda_j \left| \frac{1}{n} \sum_{i=1}^n \langle Z_i, \hat{v}_j - v_j \rangle (\langle Z_i, \boldsymbol{\theta}_0 \rangle + \epsilon_i) \right|^2 / \hat{\lambda}_j\right) \\
& + 2E\left(\sum_{j=1}^{k_n} \lambda_j \left| \frac{1}{n} \sum_{i=1}^n E_{-\hat{v}_j}[\langle Z_i, \hat{v}_j - v_j \rangle (\langle Z_i, \boldsymbol{\theta}_0 \rangle + \epsilon_i) | Z_i, W_i, \delta_i = 0] \right|^2 / \hat{\lambda}_j\right) \\
& \leq 2 \sup_{j \leq k_n} \frac{\lambda_j}{\lambda_{j+1}} * \left(\sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n E|\langle Z_i, \hat{v}_j - v_j \rangle (\langle Z_i, \boldsymbol{\theta}_0 \rangle + \epsilon_i)|^2 \right) \\
& + 2 \sup_{j \leq k_n} \frac{\lambda_j}{\lambda_{j+1}} * \left(\sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n EE_{-\hat{v}_j}[\langle Z_i, \hat{v}_j - v_j \rangle (\langle Z_i, \boldsymbol{\theta}_0 \rangle + \epsilon_i) | Z_i, W_i, \delta_i = 0] \right) \\
& \leq 2 \sup_{j \leq k_n} \frac{\lambda_j}{\lambda_{j+1}} * \left(\sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n E|\langle Z_i, \hat{v}_j - v_j \rangle (\langle Z_i, \boldsymbol{\theta}_0 \rangle + \epsilon_i)|^2 \right) \\
& + 2 \sup_{j \leq k_n} \frac{\lambda_j}{\lambda_{j+1}} * \left(\sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n EE_{-\hat{v}_j}[\langle Z_i, \hat{v}_j - v_j \rangle^2 (\langle Z_i, \boldsymbol{\theta}_0 \rangle + \epsilon_i)^2 | Z_i, W_i, \delta_i = 0] \right) \\
& \leq \text{constant} * k_n^{a-1} \sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n E[\langle Z_i, Z_i \rangle \|\hat{v}_j - v_j\|^2 (\langle Z_i, \boldsymbol{\theta}_0 \rangle + \epsilon_i)^2] \\
& + \text{constant} * k_n^{a-1} \sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n \sqrt{E\langle Z_i, Z_i \rangle^2 (\langle Z_i, \boldsymbol{\theta}_0 \rangle + \epsilon_i)^4} \sqrt{E[\|\hat{v}_j - v_j\|^4]} \\
& \leq \text{constant} * k_n^{a-1} \sum_{j=1}^{k_n} \zeta_j^{-2} n^{-1} = O(k_n^{a-1} \sum_{j=1}^{k_n} j^{3a+2}/n) = O(k_n^{4a+2} n^{-1}).
\end{aligned}$$

The last inequality holds from Lemma 1 and Lemma 2. Next, from Lemma 3, we have $B^{(6,3)} =$

$$\begin{aligned}
& \sum_{j=1}^{k_n} 2 \left(\frac{1}{\lambda_j} - \frac{1}{\hat{\lambda}_j} \right)^2 \hat{\lambda}_j^2 r_j^{*2} \lambda_j \\
& = \sum_{j=1}^{k_n} 2 \frac{(\lambda_j - \hat{\lambda}_j)^2}{\lambda_j^2} (r_j^* - \langle \boldsymbol{\theta}_0, v_j \rangle + \langle \boldsymbol{\theta}_0, v_j \rangle)^2 \lambda_j \\
& \leq \sum_{j=1}^{k_n} 4 \frac{(\lambda_j - \lambda_{j+1})^2}{\lambda_j^2} (r_j^* - \langle \boldsymbol{\theta}_0, v_j \rangle)^2 \lambda_j + \sum_{j=1}^{k_n} 4 \frac{(\lambda_j - \lambda_{j+1})^2}{\lambda_j^2} \langle \boldsymbol{\theta}_0, v_j \rangle^2 \lambda_j \\
& \leq \sum_{j=1}^{k_n} 4 (r_j^* - \langle \boldsymbol{\theta}_0, v_j \rangle)^2 \lambda_j + \sum_{j=1}^{k_n} 4 \langle \boldsymbol{\theta}_0, v_j \rangle^2 \lambda_j \\
& \leq \text{constant} \times \left[\sum_{j=1}^{k_n} \lambda_j (r_j^* - \langle \boldsymbol{\theta}_0, v_j \rangle)^2 + \sum_{j=1}^{k_n} j^{-1} j^{-2b} \right] \\
& = O_p(k_n/n) + O_p(k_n^{-2b}) = O_p(k_n/n + k_n^{-2b}).
\end{aligned}$$

Finally we have $E\left(\sum_{j=1}^{k_n} \lambda_j |r_j - r_j^*|^2\right) = A^{(6,3)} + B^{(6,3)} = O(k_n^{4a+2}/n + k_n^{-2b})$. \square

Proof of Lemma 6 (iv). The left side of the equation is equal to or less than

$$\begin{aligned}
& \sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n \frac{\mathbb{E}(\delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) \mathbb{E}(W_i \langle Z_i, v_j \rangle | Z_i, W_i, \delta_i = 0))^2}{\lambda_j} \\
\leq & \frac{1}{n} \sum_{j=1}^{k_n} \sum_{i=1}^n \mathbb{E}[\delta_i W_i \langle Z_i, v_j \rangle]^2 / \lambda_j + \frac{1}{n} \sum_{j=1}^{k_n} \sum_{i=1}^n \mathbb{E}[(1 - \delta_i) \mathbb{E}(W_i \langle Z_i, v_j \rangle | Z_i, W_i, \delta_i = 0)]^2 / \lambda_j \\
\triangleq & A^{(6,4)} + B^{(6,4)},
\end{aligned}$$

We only need to calculate the order of $A^{(6,4)}$ and $B^{(6,4)}$ separately. We have $A^{(6,4)} =$

$$\begin{aligned}
& \sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n \mathbb{E}[\delta_i W_i \langle Z_i, v_j \rangle]^2 / \lambda_j \\
= & \sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n \mathbb{E}[\delta_i W_i \xi_j^{(i)}]^2 \leq \sum_{j=1}^{k_n} \mathbb{E}[W_i \xi_j^{(i)}]^2 \leq \mathbb{E}[W_i^2],
\end{aligned}$$

and $B^{(6,4)} =$

$$\begin{aligned}
& \mathbb{E} \sum_{j=1}^{k_n} \left[\frac{1}{n} \sum_{i=1}^n (1 - \delta_i) \mathbb{E}(W_i \langle Z_i, v_j \rangle | Z_i, W_i, \delta_i = 0) \right]^2 / \lambda_j \\
\leq & \sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n \mathbb{E}[\mathbb{E}(W_i \langle Z_i, v_j \rangle | Z_i, W_i, \delta_i = 0)]^2 / \lambda_j \\
\leq & \sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n \mathbb{E}[\mathbb{E}(W_i^2 \langle Z_i, v_j \rangle^2 | Z_i, W_i, \delta_i = 0)] / \lambda_j \\
= & \sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n \mathbb{E}[W_i \xi_j^{(i)}]^2 = \sum_{j=1}^{k_n} \mathbb{E}[W_i \xi_j^{(i)}]^2 \leq \mathbb{E}[W_i^2].
\end{aligned}$$

□

Proof of Lemma 6 (v). Using Lemma 1 and Lemma 2, we have

$$\begin{aligned}
& \mathbb{E} \sup_j \zeta_j \left| \frac{1}{n} \sum_{i=1}^n \delta_i W_i \langle Z_i, \hat{v}_j - v_j \rangle \right| \\
\leq & \sqrt{\mathbb{E} \delta_i^2 W_i^2 \langle Z_i, Z_i \rangle} \sqrt{\mathbb{E} \sup_j \zeta_j^2 \|\hat{v}_j - v_j\|^2} = O_p(1/\sqrt{n}),
\end{aligned}$$

and

$$\begin{aligned}
& \mathbb{E} \sup_j \zeta_j \left| \frac{1}{n} \sum_{i=1}^n \mathbb{E}_{-\hat{v}_j} [W_i \langle Z_i, \hat{v}_j - v_j \rangle | Z_i, W_i, \delta_i = 0] \right| \\
\leq & \mathbb{E} \sup_j \frac{1}{n} \sum_{i=1}^n \mathbb{E}_{-\hat{v}_j} [\zeta_j | W_i \langle Z_i, \hat{v}_j - v_j \rangle | Z_i, W_i, \delta_i = 0] \\
\leq & \sqrt{\mathbb{E}[\mathbb{E}_{-\hat{v}_j} (W_i^2 \langle Z_i, Z_i \rangle | W_i, Z_i, \delta_i = 0)]} \sqrt{\mathbb{E} \mathbb{E}_{-\hat{v}_j} \sup_j \zeta_j \|\hat{v}_j - v_j\|^2} \\
= & O_p(1/\sqrt{n}).
\end{aligned}$$

This uses the similar technique to $A^{(6,2)}$ in the proof Lemma 6 (ii). □

Proof of Lemma 6 (vi). Denote

$$\Delta_1\left[\sum_{j=1}^{k_n} r_j(\delta_i W_i\langle Z_i, \hat{v}_j\rangle)\right] = \sum_{j=1}^{k_n} r_j(\delta_i W_i\langle Z_i, \hat{v}_j\rangle) - \sum_{j=1}^{k_n} r_j^*(\delta_i W_i\langle Z_i, v_j\rangle).$$

Then we have the following decomposition.

$$\begin{aligned} & \frac{1}{n} \sum_{i=1}^n \Delta_1\left[\sum_{j=1}^{k_n} r_j(\delta_i W_i\langle Z_i, \hat{v}_j\rangle)\right] \\ &= \sum_{j=1}^{k_n} (r_j - r_j^*) \frac{1}{n} \sum_{i=1}^n \delta_i W_i\langle Z_i, \hat{v}_j - v_j\rangle + \sum_{j=1}^{k_n} (r_j - r_j^*) \frac{1}{n} \sum_{i=1}^n \delta_i W_i\langle Z_i, v_j\rangle \\ &+ \sum_{j=1}^{k_n} (r_j^* - \langle \boldsymbol{\theta}_0, v_j\rangle) \frac{1}{n} \sum_{i=1}^n \delta_i W_i\langle Z_i, \hat{v}_j - v_j\rangle + \sum_{j=1}^{k_n} \langle \boldsymbol{\theta}_0, v_j\rangle \frac{1}{n} \sum_{i=1}^n \delta_i W_i\langle Z_i, \hat{v}_j - v_j\rangle \\ &\triangleq A^{(6,6)} + B^{(6,6)} + C^{(6,6)} + D^{(6,6)}. \end{aligned}$$

Next we bound the $A^{(6,6)}$, $B^{(6,6)}$, $C^{(6,6)}$, and $D^{(6,6)}$ separately.

From the Cauchy's inequality, by conclusions 3 and 4 of this lemma, we have $|B^{(6,6)}| \leq$

$$\begin{aligned} & \sqrt{\sum_{j=1}^{k_n} \lambda_j (r_j - r_j^*)^2} \sqrt{\sum_{j=1}^{k_n} \left[\frac{1}{n} \sum_{i=1}^n \delta_i W_i\langle Z_i, v_j\rangle\right]^2 / \lambda_j} \\ &= \sqrt{O_p(k_n^{4a+2} n^{-1} + k_n^{-2b})} O_p(1) = O_p(k_n^{2a+1} n^{-1/2} + k_n^{-b}), \end{aligned}$$

and $|A^{(6,6)}| \leq$

$$\begin{aligned} & \sum_{j=1}^{k_n} |\Delta_2(r_j)| \frac{1}{n} \sum_{i=1}^n \delta_i W_i\langle Z_i, \hat{v}_j - v_j\rangle| \\ &+ \sum_{j=1}^{k_n} \left| \frac{1}{\lambda_j} - \frac{1}{\hat{\lambda}_j} |\hat{\lambda}_j| r_j^* - \langle \boldsymbol{\theta}_0, v_j\rangle \right| \left| \frac{1}{n} \sum_{i=1}^n \delta_i W_i\langle Z_i, \hat{v}_j - v_j\rangle \right| \\ &+ \sum_{j=1}^{k_n} \left| \frac{1}{\lambda_j} - \frac{1}{\hat{\lambda}_j} |\hat{\lambda}_j| \langle \boldsymbol{\theta}_0, v_j\rangle \right| \left| \frac{1}{n} \sum_{i=1}^n \delta_i W_i\langle Z_i, \hat{v}_j - v_j\rangle \right| \\ &\triangleq A_1^{(6,6)} + A_2^{(6,6)} + A_3^{(6,6)}, \end{aligned}$$

where $\Delta_2(r_j)$ was defined in Lemma 6, (ii). To bound $A^{(6,6)}$, we only need to bound $A_1^{(6,6)}$, $A_2^{(6,6)}$ and $A_3^{(6,6)}$ separately. From Lemma 6, (ii) and Lemma 6, (v), we have

$$A_1^{(6,6)} \leq \sup_{j \leq k_n} \zeta_j \lambda_j |\Delta_2(r_j)| \sup_{j \leq k_n} \zeta_j \left| \frac{1}{n} \sum_{i=1}^n \delta_i W_i\langle Z_i, \hat{v}_j - v_j\rangle \right| \sum_{j=1}^{k_n} 1/(\lambda_j \zeta_j^2) = O_p(k_n^{4a+2} n^{-1}).$$

By the Cauchy's inequality, we have

$$\begin{aligned} & A_2^{(6,6)} + |C^{(6,6)}| \leq \left\{ \sum_{j=1}^{k_n} \frac{\left(\left| \frac{1}{\lambda_j} - \frac{1}{\hat{\lambda}_j} |\hat{\lambda}_j| / \sqrt{\lambda_j} + 1 / \sqrt{\lambda_j} \right)^2}{\zeta_j^2} \lambda_j (r_j^* - \langle \boldsymbol{\theta}_0, v_j\rangle)^2 \right. \\ & \times \left. \sum_{j=1}^{k_n} \left[\frac{\sum_{i=1}^n \zeta_j \delta_i W_i\langle Z_i, \hat{v}_j - v_j\rangle}{n} \right]^2 \right\}^{0.5}. \end{aligned}$$

Using the fact that

$$\left(\frac{1}{\lambda_j} - \frac{1}{\hat{\lambda}_j} |\hat{\lambda}_j / \sqrt{\lambda_j} + 1 / \sqrt{\lambda_j}|^2 / \zeta_j^2 \leq \text{constant} \times (j^{\frac{a}{2}})^2 j^{2a+2} = \text{constant} \times j^{3a+2}\right.$$

implied by Lemma 3, together with Lemma 6, (i), and Lemma 6, (v), we have

$$A_2^{(6,6)} + |C^{(6,6)}| = \sqrt{O_p(k_n^{3a+3}/n)O_p(1/n)} = O_p(k_n^{(3a+3)/2}/n).$$

Similarly, we have

$$A_3^{(6,6)} + |D^{(6,6)}| = \sum_{j=1}^{k_n} j^{-b} / \zeta_j O_p(1/\sqrt{n}) = O_p(\max(k_n^{a+2-b}, \log n) n^{-1/2}).$$

Therefore, $\frac{1}{n} \sum_{i=1}^n \Delta_1 [\sum_{j=1}^{k_n} r_j (\delta_i W_i \langle Z_i, \hat{v}_j \rangle)] =$

$$A_1^{(6,6)} + A_2^{(6,6)} + A_3^{(6,6)} + |B^{(6,6)}| + |C^{(6,6)}| + |D^{(6,6)}| = O_p(k_n^{2a+1} n^{-1/2} + k_n^{-b}).$$

Similarly,

$$\frac{1}{n} \sum_{i=1}^n \Delta_1 \left[\sum_{j=1}^{k_n} r_j (1 - \delta_i) E_{-\hat{v}_j} (W_i \langle Z_i, \hat{v}_j \rangle | Z_i, W_i, \delta_i = 0) \right] = O_p(k_n^{2a+1} n^{-1/2} + k_n^{-b}).$$

Finally $L_6 =$

$$\frac{1}{n} \sum_{i=1}^n \Delta_1 \left[\sum_{j=1}^{k_n} r_j (\delta_i W_i \langle Z_i, \hat{v}_j \rangle) \right] + \frac{1}{n} \sum_{i=1}^n \Delta_1 \left[\sum_{j=1}^{k_n} r_j (1 - \delta_i) E_{-\hat{v}_j} (W_i \langle Z_i, \hat{v}_j \rangle | Z_i, W_i, \delta_i = 0) \right]$$

which is $O_p(k_n^{2a+1} n^{-1/2} + k_n^{-b})$.

□

Lemma 7. Under Assumptions (A.1)–(A.4), (A.6), (A.7) and (A.9), we have $L_7 \triangleq$

$$\begin{aligned} & \frac{1}{n} \sum_{i=1}^n \left\{ \sum_{j=1}^{k_n} r_j^* [\delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) E\{\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0\}] \right\} \\ & - E\left\{ \sum_{j=1}^{k_n} r_j^* [\delta W \langle Z, v_j \rangle + (1 - \delta) E\{\langle Z, v_j \rangle W | X, V, \delta_i = 0\}] \right\} = O_p(\sqrt{k_n/n}). \end{aligned}$$

Proof. We use the following decomposition.

$$\begin{aligned} & \frac{1}{n} \sum_{i=1}^n \left\{ \sum_{j=1}^{k_n} r_j^* [\delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) E\{\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0\}] \right\} \\ & - E\left\{ \sum_{j=1}^{k_n} r_j^* [\delta W \langle Z, v_j \rangle + (1 - \delta) E\{\langle Z, v_j \rangle W | X, V, \delta_i = 0\}] \right\} \\ & \leq \frac{1}{n} \sum_{i=1}^n \left\{ \sum_{j=1}^{k_n} (r_j^* - \langle v_j, \theta_0 \rangle) [\delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) E\{\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0\}] \right\} \\ & + \sum_{j=1}^{k_n} \langle \theta_0, v_j \rangle \left[\frac{1}{n} \sum_{i=1}^n \delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) E\{\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0\} \right] \\ & - E\{W \langle Z, v_j \rangle\} \triangleq A^{(7)} + B^{(7)}. \end{aligned}$$

Denote

$$B_1^{(7)} \triangleq \mathbb{E} \sum_{j=1}^{k_n} \left(\frac{1}{n} \sum_{i=1}^n [\delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) \mathbb{E}(\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0) - \mathbb{E}\{W \langle Z, v_j \rangle\}] \right)^2.$$

Using the Jensen's inequality and the Cauchy's inequality, after algebraic calculations, we have

$$\begin{aligned} B_1^{(7)} &= \frac{1}{n} \sum_{j=1}^{k_n} \mathbb{E} [\delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) \mathbb{E}(\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0) - \mathbb{E}\{W \langle Z, v_j \rangle\}]^2 \\ &\leq \frac{3}{n} \sum_{j=1}^{k_n} [\mathbb{E}(W \langle Z, v_j \rangle)^2 + \mathbb{E}(\mathbb{E}(W \langle Z, v_j \rangle | Z_i, W_i, \delta_i = 0))^2 + \mathbb{E}^2(W \langle Z, v_j \rangle)] \\ &\leq \frac{3}{n} \sum_{j=1}^{k_n} \lambda_j (\mathbb{E}W^2 \xi_j^2 + \mathbb{E}W^2 \xi_j^2 + \mathbb{E}^2(W \xi_j)) \leq \frac{9}{n} \sum_{j=1}^{\infty} \lambda_j \sqrt{\mathbb{E}W^4 \mathbb{E}\xi_1^4} = O(1/n). \end{aligned}$$

It follows that

$$B^{(7)} \leq \sqrt{\left(\sum_{j=1}^{k_n} \langle \boldsymbol{\theta}_0, v_j \rangle^2 \right) B_1^{(7)}} \leq \|\boldsymbol{\theta}_0\| \sqrt{B_1} = O_p(1/\sqrt{n}).$$

Before we continue with $A^{(7)}$, first we denote

$$\begin{aligned} A_1^{(7)} &\triangleq \sum_{j=1}^{k_n} \frac{\left[\frac{1}{n} \sum_{i=1}^n (\delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) \mathbb{E}\{\langle Z_i, v_j \rangle W_i | Z_i, W_i, \delta_i = 0\}) \right]^2}{\lambda_j} \\ \mathbb{E}A_1^{(7)} &\leq \sum_{j=1}^{k_n} \frac{\sum_{i=1}^n \mathbb{E} \left[\delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) \mathbb{E}\{\langle Z_i, v_j \rangle W_i | Z_i, W_i, \delta_i = 0\} \right]^2}{n \lambda_j} \\ &= \sum_{j=1}^{k_n} \frac{\mathbb{E}(W_i^2 \langle Z_i, v_j \rangle^2 + \mathbb{E}(\mathbb{E}^2\{\langle Z_i, v_j \rangle W_i | Z_i, W_i, \delta_i = 0\}))}{\lambda_j} \\ &\leq \sum_{j=1}^{k_n} \frac{\mathbb{E}(W_i^2 \langle Z_i, v_j \rangle^2 + \mathbb{E}(\mathbb{E}\{\langle Z_i, v_j \rangle^2 W_i^2 | Z_i, W_i, \delta_i = 0\}))}{\lambda_j} \\ &= \sum_{j=1}^{k_n} \frac{2\mathbb{E}(W_i^2 \xi_j^2 \lambda_j)}{\lambda_j} = \sum_{j=1}^{k_n} 2\mathbb{E}(\xi_j^2 W^2) \leq 2\mathbb{E}W^2. \end{aligned}$$

Then the following equation

$$A^{(7)} \leq \sqrt{\sum_{j=1}^{k_n} \lambda_j (r_j^* - \langle v_j, \boldsymbol{\theta}_0 \rangle)^2 A_1^{(7)}} = \sqrt{O_p(k_n/n) O_p(1)} = O_p\left(\sqrt{\frac{k_n}{n}}\right),$$

holds by the conclusion 1 of Lemma 6.

Finally we have $L_7 \leq A^{(7)} + B^{(7)} = O_p(\sqrt{k_n/n})$.

□

Lemma 8. Under Assumptions (A.1)–(A.7) and (A.9), we have

$$\left\| \sum_{j=1}^{k_n} r_j \hat{v}_j - \sum_{j=1}^{k_n} \langle \boldsymbol{\theta}_0, v_j \rangle v_j \right\| = O_p(k_n^{5a/2+3/2}/\sqrt{n} + k_n^{a/2+1/2-b}).$$

Proof. First we make the decomposition of the formula.

$$\begin{aligned} & \left\| \sum_{j=1}^{k_n} r_j \hat{v}_j - \sum_{j=1}^{k_n} \langle \boldsymbol{\theta}_0, v_j \rangle v_j \right\| \\ & \leq \left\| \sum_{j=1}^{k_n} (r_j - \langle \boldsymbol{\theta}_0, v_j \rangle) (\hat{v}_j - v_j) \right\| + \left\| \sum_{j=1}^{k_n} (r_j - \langle \boldsymbol{\theta}_0, v_j \rangle) v_j \right\| \\ & + \left\| \sum_{j=1}^{k_n} \langle \boldsymbol{\theta}_0, v_j \rangle (\hat{v}_j - v_j) \right\| \triangleq A^{(8)} + B^{(8)} + C^{(8)}. \end{aligned}$$

We will discuss $A^{(8)}$, $B^{(8)}$, and $C^{(8)}$ separately. To calculate $A^{(8)}$ we define $A_1^{(8)} \triangleq \left\| \sum_{j=1}^{k_n} (r_j^* - \langle \boldsymbol{\theta}_0, v_j \rangle) (\hat{v}_j - v_j) \right\|$, and $A_2 \triangleq \left\| \sum_{j=1}^{k_n} (r_j - r_j^*) (\hat{v}_j - v_j) \right\|$. Using the conclusion of Lemma 1, Lemma 2, and Lemma 6, (i), we have $A_1^{(8)} \leq$

$$\sqrt{\sum_{j=1}^{k_n} \lambda_j (r_j^* - \langle \boldsymbol{\theta}_0, v_j \rangle)^2 / (\zeta_j^2 \lambda_j)} \sqrt{\sum_{j=1}^{k_n} \zeta_j^2 \|\hat{v}_j - v_j\|^2} = \sqrt{O_p(k_n^{3a+2+1}/n) O_p(k_n/n)}$$

which equals $O_p(k_n^{3a/2+2}/n)$; using the conclusion of Lemma 1, Lemma 2, and Lemma 6, (iii), we have $A_2^{(8)} \leq$

$$\sqrt{\sum_{j=1}^{k_n} \lambda_j (r_j - r_j^*)^2} \sqrt{\sum_{j=1}^{k_n} \|\hat{v}_j - v_j\|^2 / \lambda_j} = \sqrt{O_p(k_n^{4a+2}/n + k_n^{-2b}) O_p(k_n^{3a+3}/n)}.$$

which equals $O_p(k_n^{7a/2+5/2}/n + k_n^{3a/2+3/2-b}/\sqrt{n})$. Put them together and we get

$$A^{(8)} \leq A_1^{(8)} + A_2^{(8)} = O_p(k_n^{5a/2+3/2}/n + k_n^{a/2+1/2-b}/\sqrt{n}).$$

To calculate $B^{(8)}$ we define $B_1^{(8)} \triangleq \left\| \sum_{j=1}^{k_n} (r_j^* - \langle \boldsymbol{\theta}_0, v_j \rangle) v_j \right\|$, and $B_2^{(8)} \triangleq \left\| \sum_{j=1}^{k_n} (r_j - r_j^*) v_j \right\|$. Using the conclusion of Lemma 6, (i), we have $B_1^{(8)} \leq$

$$\sqrt{\sum_{j=1}^{k_n} \lambda_j (r_j^* - \langle \boldsymbol{\theta}_0, v_j \rangle)^2 / \lambda_j} = \sqrt{O_p(k_n^{a+1}/n)} = O_p(k_n^{(a+1)/2}/\sqrt{n});$$

using Lemma 1, 2, and 6, (iii), we have $B_2^{(8)} \leq$

$$\sqrt{\sum_{j=1}^{k_n} \lambda_j (r_j - r_j^*)^2} \sqrt{\sum_{j=1}^{k_n} \|v_j\|^2 / \lambda_j} = \sqrt{O_p(k_n^{4a+2}/n + k_n^{-2b}) O_p(k_n^{a+1})},$$

which equals $O_p(k_n^{5a/2+3/2}/\sqrt{n} + k_n^{a/2+1/2-b})$. Put them together, and we have

$$B^{(8)} \leq B_1^{(8)} + B_2^{(8)} = O_p(k_n^{5a/2+3/2}/\sqrt{n} + k_n^{a/2+1/2-b}).$$

Using Lemma 1 and Lemma 2, the following inequality holds for $C^{(8)}$.

$$\|C^{(8)}\| \leq \sum_{j=1}^{k_n} \zeta_j O_p(n^{-1/2}) |\langle \boldsymbol{\theta}_0, v_j \rangle| \leq \sum_{j=1}^{k_n} j^{a+1-b} O(n^{-1/2}) \leq O(k_n^{(a+2-b)+}/\sqrt{n}).$$

Finally, we have

$$\left\| \sum_{j=1}^{k_n} r_j \hat{v}_j - \sum_{j=1}^{k_n} \langle \boldsymbol{\theta}_0, v_j \rangle v_j \right\| \leq A^{(8)} + B^{(8)} + C^{(8)} = O_p(k_n^{5a/2+3/2}/\sqrt{n} + k_n^{a/2+1/2-b}).$$

□

Lemma 9. *Under Assumptions (A.1)–(A.9), we have*

$$\|\partial U(\beta_0)/\partial(\beta^T) - \mathfrak{J}\| = o_p(1).$$

Proof of Lemma 9. After straightforward algebraic calculations, $\partial U(\beta_0)/\partial(\beta^T) =$

$$\frac{1}{n} \sum_{i=1}^n [\delta_i W_i W_i^T + (1 - \delta_i) \mathbb{E}(W_i W_i^T | Z_i, W_i, \delta_i = 0)] - \sum_{j=1}^{k_n} R(\hat{v}_j, W) R(\hat{v}_j, W^T) / \hat{\lambda}_j,$$

where

$$R(\hat{v}_j, W) = \frac{1}{n} \sum_{i=1}^n (\delta_i \langle Z_i, \hat{v}_j \rangle W_i + (1 - \delta_i) \mathbb{E}(\langle Z_i, \hat{v}_j \rangle W_i | Z_i, W_i, \delta_i = 0)).$$

It follows that $\partial U(\beta_0)/\partial(\beta^T) - \mathfrak{J} =$

$$\begin{aligned} & \left\{ \frac{1}{n} \sum_{i=1}^n [\delta_i W_i W_i^T + (1 - \delta_i) \mathbb{E}(W_i W_i^T | Z_i, W_i, \delta_i = 0)] - \mathbb{E} W W^T \right\} \\ & - \left\{ \sum_{j=1}^{k_n} \tilde{r}_j R(\hat{v}_j, W_i^T) - \sum_{j=1}^{k_n} \tilde{r}_j^* R(v_j, W_i^T) \right\} \\ & - \left\{ \sum_{j=1}^{k_n} \tilde{r}_j^* R(v_j, W_i^T) - \sum_{j=1}^{k_n} [\mathbb{E} \langle Z, v_j \rangle W] [\mathbb{E} \langle Z, v_j \rangle W^T] / \lambda_j \right\} \\ & - \sum_{j=k_n+1}^{\infty} [\mathbb{E} \langle Z, v_j \rangle W] [\mathbb{E} \langle Z, v_j \rangle W^T] / \lambda_j \triangleq A^{(9)} + B^{(9)} + C^{(9)} + D^{(9)}, \end{aligned}$$

where

$$\begin{aligned} \tilde{r}_j &= \frac{\sum_{i=1}^n [\delta_i \tilde{M}_j(Z_i, W_i, \hat{v}_j) + (1 - \delta_i) m_{\tilde{M}_j, i, \gamma}^0(Z_i, W_i, \hat{v}_j)]}{n \hat{\lambda}_j}, \\ \tilde{r}_j^* &= \frac{\sum_{i=1}^n [\delta_i \tilde{M}_j(Z_i, W_i, v_j) + (1 - \delta_i) m_{\tilde{M}_j, i, \gamma}^0(Z_i, W_i, v_j)]}{n \lambda_j}, \end{aligned}$$

and $\tilde{M}_j(Z_i, W_i, v_j) = \langle Z_i, v_j \rangle W_i$. Note that $\|A^{(9)}\| = o_p(1)$ holds by law of large numbers; $B^{(9)} =$

$$\sum_{j=1}^{k_n} (\tilde{r}_j - \tilde{r}_j^*) \left[\frac{1}{n} \sum_{i=1}^n \langle Z_i, (\hat{v}_j - v_j) \rangle W_i^T \right] + \sum_{j=1}^{k_n} (\tilde{r}_j - \tilde{r}_j^*) \left[\frac{1}{n} \sum_{i=1}^n \langle Z_i, v_j \rangle W_i^T \right] + \sum_{j=1}^{k_n} \tilde{r}_j^* \left[\frac{1}{n} \sum_{i=1}^n \langle Z_i, \hat{v}_j - v_j \rangle W_i^T \right],$$

which equals $O_p(k_n^{2a+1} n^{-1/2} + k_n^{a-b-1})$ using the technique similar to the proof of Lemma 6, (vi); similar to the proof of Lemma 7, $\|C\| = O_p(\sqrt{k_n/n})$; $\|D^{(9)}\| = o(1)$ since

$$\sum_{j=1}^{\infty} [\mathbb{E} \langle Z, v_j \rangle W] [\mathbb{E} \langle Z, v_j \rangle W^T] / \lambda_j = \sum_{j=1}^{\infty} \mathbb{E} \xi_j W \mathbb{E} \xi_j W^T \leq \mathbb{E} W W^T$$

holds by Assumption (A.8). In conclusion,

$$\|\partial U(\beta_0)/\partial(\beta^T) - \mathfrak{J}\| \leq \|A^{(9)}\| + \|B^{(9)}\| + \|C^{(9)}\| + \|D^{(9)}\| = o_p(1).$$

□

Proof of Theorem 1. First we denote $e \triangleq \partial U(\beta_1)/\partial \beta_1^T - \mathfrak{J}$. From Lemma 4–Lemma 7, we have

$$U(\beta_{1,0}) = L_5 - L_4 + L_6 + L_7 = O_p(k_n^{2a+1}n^{-1/2} + k_n^{1/2-b}),$$

where the ‘ L_i ’ is defined in Lemma i , $i = 4, 5, 6, 7$. Then from $\|e\| \rightarrow 0$ of Lemma 9, together with Equation (1.5) of Stewart (1969), we have $\|\hat{\beta}_{1,0} - \beta_{1,0}\| =$

$$\begin{aligned} & \left\| \left(\frac{\partial U(\beta_1)}{\partial \beta_1^T} \right)^{-1} U(\beta_{1,0}) \right\| \leq \left\| \left(\frac{\partial U(\beta_1)}{\partial \beta_1^T} \right)^{-1} \right\| \|U(\beta_{1,0})\| = \|(\mathfrak{J} + e)^{-1}\| \|U(\beta_{1,0})\| \\ & \leq [\|(\mathfrak{J} + e)^{-1} - \mathfrak{J}^{-1}\| + \|\mathfrak{J}^{-1}\|] \|U(\beta_{1,0})\| \leq [\|(\mathfrak{J} + e)^{-1} - \mathfrak{J}^{-1}\| \|\mathfrak{J}^{-1}\| + 1] \|U(\beta_{1,0})\| \|\mathfrak{J}^{-1}\| \\ & \leq \left[\frac{\|\mathfrak{J}^{-1}\| \|e\|}{1 - \|\mathfrak{J}^{-1}\| \|e\|} + 1 \right] \|U(\beta_{1,0})\| \|\mathfrak{J}^{-1}\| \leq \frac{\|U(\beta_{1,0})\| \|\mathfrak{J}^{-1}\|}{1 - \|\mathfrak{J}^{-1}\| \|e\|} = O_p(\|U(\beta_{1,0})\|) \end{aligned}$$

which also equals $O_p(k_n^{2a+1}n^{-1/2} + k_n^{1/2-b})$. Then the first part of the theorem is proved.

For the second part of the theorem, we make the following decompositions. $\theta(\hat{\beta}_1) - \theta_0 =$

$$\begin{aligned} & \left\{ \sum_{j=1}^{k_n} (r_j(\hat{\beta}_1) - r_j(\beta_{1,0})) \hat{v}_j \right\} + \left\{ \sum_{j=1}^{k_n} r_j(\beta_{1,0}) \hat{v}_j - \sum_{j=1}^{k_n} \langle \theta_0, v_j \rangle v_j \right\} + \left\{ - \sum_{j=k_n+1}^{\infty} \langle \theta_0, v_j \rangle v_j \right\} \\ & \triangleq A^{\mathbf{T}1} + B^{\mathbf{T}1} + C^{\mathbf{T}1}. \end{aligned}$$

From Lemma 8, we have $\|B^{\mathbf{T}1}\| = O_p(k_n^{5a/2+3/2}/\sqrt{n} + k_n^{a/2+1/2-b})$. Therefore we only need to calculate the three terms $A^{\mathbf{T}1}$, $B^{\mathbf{T}1}$ and $C^{\mathbf{T}1}$. Note that

$$\|C^{\mathbf{T}1}\| = \sqrt{\langle C, C \rangle} = \sqrt{\sum_{j=k_n+1}^{\infty} \langle \theta_0, v_j \rangle^2} = O\left(\sqrt{\sum_{j>k_n} j^{-2b}}\right) = O_p(k_n^{1/2-b}),$$

and $A^{\mathbf{T}1}$ has the following decompositions, $A^{\mathbf{T}1} = (A_1^{\mathbf{T}1} + A_2^{\mathbf{T}1} + A_3^{\mathbf{T}1} + A_4^{\mathbf{T}1})(\hat{\beta}_1 - \beta_{1,0})$, where

$$\begin{aligned} A_1^{\mathbf{T}1} &= \sum_{j=1}^{k_n} (\tilde{r}_j - E\langle Z, v_j \rangle W/\lambda_j) (\hat{v}_j - v_j); \quad A_2^{\mathbf{T}1} = \sum_{j=1}^{k_n} (\tilde{r}_j - E\langle Z, v_j \rangle W/\lambda_j) v_j; \\ A_3^{\mathbf{T}1} &= \sum_{j=1}^{k_n} E\langle Z, v_j \rangle W/\lambda_j (\hat{v}_j - v_j); \quad A_4^{\mathbf{T}1} = \sum_{j=1}^{k_n} E\langle Z, v_j \rangle W/\lambda_j v_j. \end{aligned}$$

In the following we will calculate the four terms $A_1^{\mathbf{T}1}$, $A_2^{\mathbf{T}1}$, $A_3^{\mathbf{T}1}$ and $A_4^{\mathbf{T}1}$. We have $\|A_3^{\mathbf{T}1}\| \leq$

$$\sqrt{E \sum_{j=1}^{k_n} \langle Z, v_j \rangle^2 W^2 / \lambda_j^2 \sum_{j=1}^{k_n} \|\hat{v}_j - v_j\|^2} = O_p(k_n^{(a+1)/2}) \sqrt{O_p\left(\sum_{j=1}^{k_n} 1/\zeta_j^2\right) \frac{1}{n}} = O_p(k_n^{(3a+4)/2}/\sqrt{n})$$

using Lemma 1, 2 and the Cauchy’s inequality; we have $\|A_4^{\mathbf{T}1}\| \leq$

$$\sqrt{E \sum_{j=1}^{k_n} \langle Z, v_j \rangle^2 W^2 / \lambda_j^2} = \sqrt{\sum_{j=1}^{k_n} E(\xi_j^2 W^2) / \lambda_j} = O_p(k_n^{(a+1)/2})$$

by Assumption 2; similar to Lemma 6, (i), we have

$$\sum_{j=1}^{k_n} j^x \lambda_j (\tilde{r}_j^* - E\langle Z, v_j \rangle W/\lambda_j)^2 = O_p(k_n^{1+x}/n), \text{ for any } x \neq -1,$$

so that the following equation holds,

$$\|A_{21}^{\mathbf{T}1}\| \triangleq \sqrt{\sum_{j=1}^{k_n} [\tilde{r}_j^* - \mathbf{E}\langle Z, v_j \rangle W / \lambda_j]^2} = O_p(k_n^{(a+1)/2});$$

similar to Lemma 6, (iii), we have

$$\sum_{j=1}^{k_n} \lambda_j |\tilde{r}_j - \tilde{r}_j^*|^2 = O_p(k_n^{4a+2} n^{-1} + k_n^{-2b}),$$

so that the following equation holds,

$$\begin{aligned} \|A_{22}^{\mathbf{T}1}\| &\triangleq \sqrt{\sum_{j=1}^{k_n} [\tilde{r}_j - \tilde{r}_j^*]^2} \leq \sqrt{\sum_{j=1}^{k_n} \lambda_j [\tilde{r}_j - \tilde{r}_j^*]^2 / \lambda_{k_n}} \\ &= \sqrt{O_p(k_n^{4a+2} n^{-1} + k_n^{-2b}) O_p(k_n^a)} = O_p(k_n^{5/2a+1} n^{-1/2} + k_n^{a/2-b}); \end{aligned}$$

consequently, we have $\|A_2^{\mathbf{T}1}\| \leq \|A_{21}^{\mathbf{T}1}\| + \|A_{22}^{\mathbf{T}1}\| = O_p(k_n^{(a+1)/2})$. Similar to the derivation of $\|A_3^{\mathbf{T}1}\|$, we can prove that $\|A_1^{\mathbf{T}1}\| = O_p(\|A_2^{\mathbf{T}1}\|)$.

In all,

$$\begin{aligned} \|A^{\mathbf{T}1}\| &\leq \|A_1^{\mathbf{T}1}\| + \|A_2^{\mathbf{T}1}\| + \|A_3^{\mathbf{T}1}\| + \|A_4^{\mathbf{T}1}\| \|\hat{\boldsymbol{\beta}}_1 - \boldsymbol{\beta}_{1,0}\| \\ &= O_p(k_n^{(a+1)/2}) O_p(k_n^{2a+1} n^{-1/2} + k_n^{1/2-b}) = O_p(k_n^{5/2a+3/2} n^{-1/2} + k_n^{1+a/2-b}). \end{aligned}$$

and finally, we get

$$\begin{aligned} \|\boldsymbol{\theta}(\hat{\boldsymbol{\beta}}_1) - \boldsymbol{\theta}_0\| &\leq \|A^{\mathbf{T}1}\| + \|B^{\mathbf{T}1}\| + \|C^{\mathbf{T}1}\| \\ &= O_p(k_n^{5/2a+3/2} n^{-1/2} + k_n^{1+a/2-b}). \end{aligned}$$

□

Lemma 10. *Under Assumption A.1–A.9 and B.1–B.5, we have*

(i)

$$F(\langle Z_l, \boldsymbol{\theta}_0 \rangle, G(\langle Z_l, W_l \rangle), W_l) = \mathbf{E}[\langle Z_l, \boldsymbol{\theta}_0 \rangle \delta_l \exp(\gamma_0 Y_l) | Z_l, W_l];$$

(ii) *for any $C_1 > 0$, there exists a constant $C_2 > 0$, such that*

$$\sup_{\sum_{i=1}^3 \|x_i\| \leq C_1} \max_{i=1,2,3} \left| \frac{\partial F(x_1, x_2, x_3)}{\partial x_i} \right| \leq C_2, \quad \sup_{\sum_{i=1}^3 \|x_i\| \leq C_1} F(x_1, x_2, x_3) \leq C_2,$$

holds.

Proof of Lemma 10 (i) By algebraic calculations, we have

$$\begin{aligned} &\mathbf{E}[\langle Z_l, \boldsymbol{\theta}_0 \rangle \delta_l \exp(\gamma_0 Y_l) | Z_l, W_l] \\ &= \langle Z_l, \boldsymbol{\theta}_0 \rangle \mathbf{E}[\Pr(\delta_l = 1 | Z_l, W_l, Y_l) \exp(\gamma_0 Y_l) | Z_l, W_l] \\ &= \langle Z_l, \boldsymbol{\theta}_0 \rangle \mathbf{E}\left\{ \frac{\exp(G(Z_l, W_l))}{1 + \exp(\langle g, Z_l \rangle + \langle \boldsymbol{\beta}_{2,0}, W_l \rangle + \phi_0 Y_l)} \middle| Z_l, W_l \right\} \\ &= \int \frac{\langle Z_l, \boldsymbol{\theta}_0 \rangle \exp(G(Z_l, W_l))}{1 + \exp(G(Z_l, W_l) + \phi_0 y)} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{(y - \langle Z_l, \boldsymbol{\theta}_0 \rangle - \boldsymbol{\beta}_{1,0}^T W_l)^2}{2\sigma^2}\right) dy, \end{aligned}$$

which is a measurable function of $\langle Z_l, \boldsymbol{\theta}_0 \rangle, G(Z_l, W_l)$ and W .

□

Proof of Lemma 10 (ii) Continue with Lemma 10, (i), and we obtain the explicit form of the function F in the following.

$$F(x_1, x_2, x_3) = \frac{1}{\sqrt{2\pi}} \int \frac{x_1 \exp(x_2)}{1 + \exp(x_2 + \boldsymbol{\beta}_{2,0}^T x_3 + \phi_0 y)} \exp\left(-\frac{(y - x_1 - \boldsymbol{\beta}_{1,0}^T x_3)^2}{2\sigma^2}\right) dy,$$

and

$$\begin{aligned} \frac{\partial F(x_1, x_2, x_3)}{\partial x_1} &= \frac{1}{\sqrt{2\pi}} \int \frac{\exp(x_2)}{1 + \exp(x_2 + \phi_0 y)} \exp\left(-\frac{(y - x_1 - \boldsymbol{\beta}_{1,0}^T x_3)^2}{2\sigma^2}\right) dy \\ &+ \frac{1}{\sqrt{2\pi}} \int \frac{x_1 \exp(x_2)}{1 + \exp(x_2 + \phi_0 y)} \exp\left(-\frac{(y - x_1 - \boldsymbol{\beta}_{1,0}^T x_3)^2}{2\sigma^2}\right) \frac{y - x_1 - \boldsymbol{\beta}_{1,0}^T x_3}{\sigma^2} dy. \end{aligned}$$

Note that both F and $\partial F/\partial x_1$ are continuous functions of (x_1, x_2, x_3) within the compact set

$$\{(x_1, x_2, x_3) \mid \sum_{i=1}^3 \|x_i\| \leq C_1\}.$$

Therefore, there exists $C_{2,1} > 0, C_{2,0} > 0$ such that

$$\sup_{\sum_{i=1}^3 \|x_i\| \leq C_1} \left| \frac{\partial F(x_1, x_2, x_3)}{\partial x_1} \right| \leq C_{2,1}, \quad \sup_{\sum_{i=1}^3 \|x_i\| \leq C_1} F(x_1, x_2, x_3) \leq C_{2,0}$$

Similarly, there exist constants $C_{2,i} > 0, i = 2, 3$ such that

$$\sup_{\sum_{i=1}^3 \|x_i\| \leq C_1} \left| \frac{\partial F(x_1, x_2, x_3)}{\partial x_i} \right| \leq C_{2,i},$$

and then we have

$$\sup_{\sum_{i=1}^3 \|x_i\| \leq C_1} \max_i \left| \frac{\partial F(x_1, x_2, x_3)}{\partial x_i} \right| \leq \max_{i \leq 3} (C_{2,i}),$$

which completes the proof. □

Lemma 11. Under Assumption A.1–A.9 and B.1–B.5, we have

(i) Given a constant function $z \in \mathbb{H}$, a constant vector $x \in \mathbb{R}^p$, and a constant $w_0 \in (0, 1)$, there exist constants $0 \leq c_1 \leq c_2 < \infty$ such that

$$c_1 \psi_{z,x}(h) \leq \mathbb{E}[K_h(w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z - z, v_{j_1} \rangle^2 + (1 - w_0) \|W - x\|})] \leq c_2 \phi_x(h/(1 - w_0)),$$

where $\psi_{z,x}(h)$ is defined in Assumption (B.4) and $\phi_x(h) \triangleq \Pr(W \in \{\tilde{x} \mid \|\tilde{x} - x\| \leq h\})$.

(ii) The following two inequalities

$$\begin{aligned} & \sup_{z,x} \mathbb{E} \left\{ \left[\frac{1/n \sum_{l=1}^n K_h^{(l)}(w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z_l - Z_i, v_{j_1} \rangle^2 + (1 - w_0) \|W_l - W_i\|})}{\mathbb{E}[K_h(w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z - Z_i, v_{j_1} \rangle^2 + (1 - w_0) \|W - W_i\|}) \mid Z_i, W_i]} \right]^2 \mid Z_i = z, W_i = x \right\} \\ & \triangleq L_{11}^{(1)} \leq \text{constant} \times \left(h^2 + \frac{1}{n\psi(h)} \right), \end{aligned}$$

and

$$\begin{aligned} & \sup_{z,x} \mathbb{E} \left\{ \left[\frac{1/n \sum_{l=1}^n \tilde{K}_h^{(l)}(w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z_l - Z_i, v_{j_1} \rangle^2} + (1-w_0) \|W_l - W_i\|})}{\mathbb{E}[K_h(w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z - Z_i, v_{j_1} \rangle^2} + (1-w_0) \|W - W_i\|) | Z_i, W_i]} \right]^2 \right. \\ & \left. - \mathbb{E}(\delta_i \exp(\gamma Y_i) | Z_i, W_i) \right]^2 | Z_i = z, W_i = x \} \\ & \triangleq L_{11}^{(2)} \leq \text{constant} \times \left(h^2 + \frac{1}{n\psi(h)} \right) \end{aligned}$$

hold.

Proof of Lemma 11 (i). From Assumption (B.3) and the definition of type I kernel in Martinez C. A. (2013), we have

$$\begin{aligned} & \mathbb{E}[K_h(w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z - z, v_{j_1} \rangle^2} + (1-w_0) \|W - x\|)] \\ & \geq c_1 \Pr((Z, W) \in \{(\tilde{z}, \tilde{x}) \mid w_0 \sqrt{\sum_{j=1}^{k_n} \langle \tilde{z} - z, v_j \rangle^2} + (1-w_0) \|\tilde{x} - x\| \leq h\}) \\ & \geq c_1 \Pr((Z, W) \in \{(\tilde{z}, \tilde{x}) \mid \sqrt{\sum_{j=1}^{k_n} \langle \tilde{z} - z, v_j \rangle^2} \leq h, \|\tilde{x} - x\| \leq h\}) \\ & \geq c_1 \Pr((Z, W) \in \{(\tilde{z}, \tilde{x}) \mid \|\tilde{z} - z\| \leq h, \|\tilde{x} - x\| \leq h\}) = c_1 \psi_{z,x}(h), \end{aligned}$$

and

$$\begin{aligned} & \mathbb{E}[K_h(w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z - z, v_{j_1} \rangle^2} + (1-w_0) \|W - x\|)] \\ & \leq c_2 \Pr((Z, W) \in \{(\tilde{z}, \tilde{x}) \mid w_0 \sqrt{\sum_{j=1}^{k_n} \langle \tilde{z} - z, v_j \rangle^2} + (1-w_0) \|\tilde{x} - x\| \leq h\}) \\ & \leq c_2 \Pr((Z, W) \in \{(\tilde{z}, \tilde{x}) \mid (1-w_0) \|\tilde{x} - x\| \leq h\}) = c_2 \phi_x\left(\frac{h}{1-w_0}\right), \end{aligned}$$

which complete the proof. \square

Proof of Lemma 11 (ii). Before the proof, note that from Assumption (B.3) the kernel function $K(\cdot)$ satisfies,

$$\int K(t) dt = 1; \int tK(t) dt = 0; \int K^2(t) dt \leq \text{constant}.$$

We divide the proof into two parts. In Part 1, we calculate the bias while in Part 2 the variance is calculated.

Part 1. For simplicity, denote

$$\begin{aligned} D_l^{(i)} &= w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z_l - Z_i, v_{j_1} \rangle^2} + (1-w_0) \|W_l - W_i\|, \\ R_0^{(i)} &= \mathbb{E}[K_h(w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z - Z_i, v_{j_1} \rangle^2} + (1-w_0) \|W - W_i\|) | Z_i, W_i], \end{aligned}$$

and

$$E_0^{(i)} = E(\langle Z_i, \boldsymbol{\theta}_0 \rangle \delta_i \exp(\gamma_0 Y_i) | Z_i, W_i).$$

In this step we calculate the bias $A^{(11)}(Z_i, W_i)$ first. Using Lemma 10, (i), Assumption B.3, and Lemma 11, (i), we have

$$\begin{aligned} A^{(11)}(Z_i, W_i) &\triangleq \text{Bias}\left(\frac{1/n \sum_{l=1}^n K_h^{(l)}(D_l^{(i)})}{R_0^{(i)}} | Z_i, W_i\right) = E\left[\frac{1/n \sum_{l=1}^n K_h^{(l)}(D_l^{(i)})}{R_0^{(i)}} | Z_i, W_i\right] - E_0^{(i)} \\ &= E\left[\frac{\frac{1}{n} \sum_{l=1}^n K_h(D_l^{(i)}) \delta_l \exp(\gamma_0 Y_l) \langle Z_l, \boldsymbol{\theta}_0 \rangle}{R_0^{(i)}} | Z_i, W_i\right] - E_0^{(i)} \\ &= E\left[\left\{1/n \sum_{l=1}^n K_h(D_l^{(i)}) E[\delta_l \exp(\gamma_0 Y_l) \langle Z_l, \boldsymbol{\theta}_0 \rangle | Z_i, W_i, Z_l, W_l] / R_0^{(i)}\right\} - E_0^{(i)}\right] \\ &= E\left[1/n \sum_{l=1}^n K_h(D_l^{(i)}) F(\langle Z_l, \boldsymbol{\theta}_0 \rangle, G(Z_l, W_l), W_l) | Z_i, W_i\right] / R_0^{(i)} - E_0^{(i)} \\ &= E\left[1/n \sum_{l=1}^n K_h(D_l^{(i)}) 1_{D_l^{(i)} \leq h} F(\langle Z_l, \boldsymbol{\theta}_0 \rangle, G(Z_l, W_l), W_l) | Z_i, W_i\right] / R_0^{(i)} - E_0^{(i)} \\ &\leq E\left[1/n \sum_{l=1}^n K_h(D_l^{(i)}) 1_{\|Z_l - Z_i\| \leq \frac{h}{w_0}} 1_{\|W_l - W_i\| \leq \frac{h}{1-w_0}} F(\langle Z_l, \boldsymbol{\theta}_0 \rangle, G(Z_l, W_l), W_l) | Z_i, W_i\right] / R_0^{(i)} - E_0^{(i)} \\ &\leq E\left[1/n \sum_{l=1}^n K_h(D_l^{(i)}) | Z_i, W_i\right] [F(\langle Z_i, \boldsymbol{\theta}_0 \rangle, G(Z_i, W_i), W_i) + \text{constant} \times h] / R_0^{(i)} - E_0^{(i)} \\ &= R_0^{(i)} (E_0^{(i)} + \text{constant} \times h) / R_0^{(i)} - E_0^{(i)} \end{aligned} \tag{S1.3}$$

which equals $\text{constant} * h$. Next we prove that this constant exists uniformly for $Z_i = z$ and $W_i = w$; in other words, there exists constant C such that

$$\Pr\left(\sup_{Z_i, W_i} A^{(11)}(Z_i, W_i) \leq Ch\right) = 1.$$

From Assumption (B.1), we have $|\langle Z_i, \boldsymbol{\theta}_0 \rangle| \leq \|Z_i\| \|\boldsymbol{\theta}_0\| \leq \text{constant}$ and $\|W_l\| \leq \text{constant}$; from Assumption (B.2), we have $|G(Z, W)| \leq G_1^{-1}(\|Z\| + \|W\|) \leq \text{constant}$. It follows from Lemma 10, (ii) that there exist constant C_2 , such that

$$\begin{aligned} &|F(\langle Z_l, \boldsymbol{\theta}_0 \rangle, G(Z_l, W_l), W_l) - F(\langle Z_i, \boldsymbol{\theta}_0 \rangle, G(Z_i, W_i), W_i)| \\ &\leq |F(\langle Z_l, \boldsymbol{\theta}_0 \rangle, G(Z_l, W_l), W_l) - F(\langle Z_i, \boldsymbol{\theta}_0 \rangle, G(Z_l, W_l), W_l)| \\ &+ |F(\langle Z_i, \boldsymbol{\theta}_0 \rangle, G(Z_l, W_l), W_l) - F(\langle Z_i, \boldsymbol{\theta}_0 \rangle, G(Z_i, W_i), W_l)| \\ &+ |F(\langle Z_i, \boldsymbol{\theta}_0 \rangle, G(Z_i, W_i), W_l) - F(\langle Z_i, \boldsymbol{\theta}_0 \rangle, G(Z_i, W_i), W_i)| \\ &\leq C_2[|\langle Z_l - Z_i, \boldsymbol{\theta}_0 \rangle| + |G(Z_l, W_l) - G(Z_i, W_i)| + \|W_l - W_i\|] \\ &\leq \text{constant} * [|\langle Z_l - Z_i, \boldsymbol{\theta}_0 \rangle| + \|Z_l - Z_i\| + \|W_l - W_i\| + \|W_l - W_i\|], \end{aligned}$$

The last inequality holds from the Lipschitz's condition in Assumption B.2, and the constant here is irrelevant to Z_l, Z_i, W_l and W_i . So that

$$\begin{aligned} &1_{\|Z_l - Z_i\| \leq \frac{h}{w_0}} 1_{\|W_l - W_i\| \leq \frac{h}{1-w_0}} F(\langle Z_l, \boldsymbol{\theta}_0 \rangle, G(Z_l, W_l), W_l) \\ &\leq 1_{\|Z_l - Z_i\| \leq \frac{h}{w_0}} 1_{\|W_l - W_i\| \leq \frac{h}{1-w_0}} [F(\langle Z_i, \boldsymbol{\theta}_0 \rangle, G(Z_i, W_i), W_i) + \text{constant} \times h], \end{aligned}$$

and the constant here is irrelevant to Z_l, Z_i, W_l and W_i . This illustrate the constant in (??) is irrelevant to Z_l, Z_i, W_l and W_i . It follows that

$$\Pr\left(\sup_{Z_i, W_i} A^{(11)}(Z_i, W_i) \leq Ch\right) = 1.$$

Part 2. First we calculate

$$B^{(11)}(Z_i, W_i) \triangleq \mathbb{E}[[K_h^{(l)}(D_l^{(i)})]^2 | Z_i, W_i] / [R_0^{(i)}]^2,$$

for $l \neq i$. Using Lemma 10, (ii), we have $B^{(11)}(Z_i, W_i) =$

$$\begin{aligned} & \frac{\mathbb{E}[K_h^2(D_l^{(i)}) \delta_l [\exp(\gamma_0 Y_l) \langle Z_l, \boldsymbol{\theta}_0 \rangle]^2 | Z_i, W_i]}{[R_0^{(i)}]^2} \\ &= \mathbb{E}[K_h^2(D_l^{(i)}) \langle Z_l, \boldsymbol{\theta}_0 \rangle \times \mathbb{E}[\delta_l \exp(2\gamma_0 Y_l) \langle Z_l, \boldsymbol{\theta}_0 \rangle | Z_i, W_i, Z_l, W_l] | Z_i, W_i] / [R_0^{(i)}]^2 \\ &= \mathbb{E}[K_h^2(D_l^{(i)}) \exp(\gamma_0 Y_l) \langle Z_l, \boldsymbol{\theta}_0 \rangle F(\langle Z_l, \boldsymbol{\theta}_0 \rangle, \langle Z_l, g \rangle, W_l; 2\gamma_0) | Z_i, W_i] / [R_0^{(i)}]^2 \\ &= \text{constant} \times \mathbb{E}[K_h^2(D_l) | Z_i, W_i] / [R_0^{(i)}]^2 \\ &\leq \text{constant} \times \mathbb{E}[K_h(D_l^{(i)}) | Z_i, W_i] / [R_0^{(i)}]^2 = \text{constant} \times \frac{1}{R_0^{(i)}}. \end{aligned}$$

The last inequality is from Assumption (B.3).

Then for $l_0 \neq i$, we have

$$\begin{aligned} & \text{Var}[1/n \sum_{l=1}^n [K_h^{(l)}(D_l^{(i)})] / R_0^{(i)} | Z_i, W_i] \\ &\leq \frac{1}{n^2 [R_0^{(i)}]^2} n \text{Var}[[K_h^{(l_0)}(D_{l_0}^{(i)})] | Z_i, W_i] = \frac{1}{n} \text{Var}[[K_h^{(l_0)}(D_{l_0}^{(i)})] / R_0^{(i)} | Z_i, W_i] \\ &\leq \frac{1}{n} \mathbb{E}[K_h^2(D_{l_0}^{(i)}) / [R_0^{(i)}]^2 | Z_i, W_i] = O_p\left(\frac{1}{n R_0^{(i)}}\right), \end{aligned}$$

where the O_p, o_p term and the relevant constant hold uniformly with respect to Z_i, W_i by Assumption (B.1).

Finally, we have

$$\begin{aligned} L_{11}^{(1)} &\leq \sup_{Z_i, W_i} [A^{(11)}(Z_i, W_i)]^2 + \sup_{Z_i, W_i} \text{Var}\left(\frac{1/n \sum_{l=1}^n K_h^{(l)}(D_l)}{R_0} | Z_i, W_i\right) \\ &\leq \text{constant} \times \left(h^2 + \sup_{Z_i, W_i} \frac{1}{n \psi_{Z_i, X_i}(h)}\right) \\ &\leq \text{constant} \times \left(h^2 + \frac{1}{n \psi(h)}\right). \end{aligned}$$

It can be proved in the same way that

$$L_{11}^{(2)} \leq \text{constant} \times \left(h^2 + \frac{1}{n \psi(h)}\right).$$

□

Lemma 12. Under Assumption A.1–A.9 and B.1–B.5, we have

(i)

$$\sup_{j \leq k_n} \zeta_j \lambda_j |\Delta_2(\hat{r}_j)| = O_p(k_n^{a-1} / \sqrt{n}),$$

where

$$\Delta_2(\hat{r}_j) = \hat{r}_j - \frac{\sum_{i=1}^n [\delta_i M_j(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_{1,0}) + (1 - \delta_i) \hat{m}_{M_j, i, \gamma}(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_{1,0})]}{n \hat{\lambda}_j},$$

(ii)

$$\sum_{j=1}^{k_n} \lambda_j |\hat{r}_j - \hat{r}_j^*|^2 = O_p(k_n^{4a+2}n^{-1} + k_n^{-2b});$$

(iii)

$$\sum_{j=1}^{\infty} \lambda_j^2 |\hat{r}_j^* - r_j^*|^2 = O_p[h^2 + \frac{1}{n\psi(h)}].$$

(iv) $L_{12} \triangleq$

$$\begin{aligned} & \frac{1}{n} \sum_{i=1}^n \Delta_1 \left\{ \sum_{j=1}^{k_n} \hat{r}_j [\delta_i W_i \langle Z_i, \hat{v}_j \rangle + \sum_{l=0}^n w_{l,0}(Z_i, W_i; \gamma) \langle Z_l, \hat{v}_j \rangle W_l] \right\} \\ &= O_p(k_n^{2a+1}n^{-1/2} + k_n^{-b} + k_n^{(4a+3)/2}n^{-1/2}[h + \frac{1}{\sqrt{n\psi(h)}}]), \end{aligned}$$

where

$$\begin{aligned} & \Delta_1 \left\{ \sum_{j=1}^{k_n} \hat{r}_j [\delta_i W_i \langle Z_i, \hat{v}_j \rangle + \sum_{l=0}^n w_{l,0}(Z_i, W_i; \gamma) \langle Z_l, \hat{v}_j \rangle W_l] \right\} \\ &= \sum_{j=1}^{k_n} \hat{r}_j [\delta_i W_i \langle Z_i, \hat{v}_j \rangle + \sum_{l=0}^n w_{l,0}(Z_i, W_i; \gamma) \langle Z_l, \hat{v}_j \rangle W_l] \\ &- \sum_{j=1}^{k_n} \hat{r}_j^* [\delta_i W_i \langle Z_i, v_j \rangle + \sum_{l=0}^n w_{l,0}(Z_i, W_i; \gamma) \langle Z_l, v_j \rangle W_l]. \end{aligned}$$

Proof of Lemma 12 (i). The proof is exactly the same as Lemma 6, (ii) except that it uses $A_2^{(12,1)}$ to replace $A_2^{(6,2)}$, where $A_2^{(12,1)}$ is expressed in the following (here $w_{l,0}^{(i)} \triangleq w_{l,0}(Z_i, W_i; \gamma)$ which is defined by (2.7)). $A_2^{(12,1)} \triangleq$

$$\begin{aligned} & \mathbb{E}[\sup_{j \leq k_n} \zeta_j \sum_{l=1}^n w_{l,0}^{(i)} \langle Z_l, \hat{v}_j - v_j \rangle (\langle Z_l, \boldsymbol{\theta}_0 \rangle + \epsilon_l)] \\ &= \sum_{l_0=1}^n \Pr(\arg \max_{l \leq n} w_{l,0}^{(i)} = l_0) \mathbb{E}[\sup_{j \leq k_n} \zeta_j \langle Z_{l_0}, \hat{v}_j - v_j \rangle (\langle Z_{l_0}, \boldsymbol{\theta}_0 \rangle + \epsilon_{l_0}) | \arg \max_{l \leq n} w_{l,0}^{(i)} = l_0] \\ &\leq \sup_{l_0 \leq n} \mathbb{E}[\sup_{j \leq k_n} \zeta_j \langle Z_{l_0}, \hat{v}_j - v_j \rangle (\langle Z_{l_0}, \boldsymbol{\theta}_0 \rangle + \epsilon_{l_0}) | \arg \max_{l \leq n} w_{l,0}^{(i)} = l_0] \\ &= \mathbb{E}[\sup_{j \leq k_n} \zeta_j \langle Z_{l_0}, \hat{v}_j - v_j \rangle (\langle Z_{l_0}, \boldsymbol{\theta}_0 \rangle + \epsilon_{l_0}) | \arg \max_{l \leq n} w_{l,0}^{(i)} = l_0] \\ &\leq \sqrt{\mathbb{E}[\langle Z, Z \rangle (\langle Z, \boldsymbol{\theta}_0 \rangle + \epsilon)^2 | \arg \max_{l \leq n} w_{l,0}^{(i)} = l_0]} \\ &\quad \sqrt{\mathbb{E}[\sup_j \zeta_j^2 \|\hat{v}_j - v_j\|^2 | \arg \max_{l \leq n} w_{l,0}^{(i)} = l_0]} \\ &\leq \text{constant} \times O_p(n^{-1/2}). \end{aligned}$$

□

Proof of Lemma 12 (ii). The proof is exactly the same as Lemma 6, (iii), except using

$$A^{(12,2)} = \sum_{j=1}^{k_n} 2\lambda_j \Delta_2(\hat{r}_j)^2$$

to replace $A^{(6,3)}$, where its expectation is calculated in the following (here $w_{l,0}^{(i)} \triangleq w_{l,0}(Z_i, W_i; \gamma)$ which is defined by (2.7)). Since $w_{l,0}^{(i)}$ is positive, we have $\sum_{l=1}^n (w_{l,0}^{(i)})^2 \leq (\sum_{l=1}^n w_{l,0}^{(i)})^2 = 1$, and it follows that $EA^{(12,2)} =$

$$\begin{aligned}
& 2E\left(\sum_{j=1}^{k_n} \lambda_j \left| \frac{1}{n} \sum_{i=1}^n \langle Z_i, \hat{v}_j - v_j \rangle (\langle Z_i, \boldsymbol{\theta}_0 \rangle + \epsilon_i) \right|^2 / \hat{\lambda}_j\right) \\
& + 2E\left(\sum_{j=1}^{k_n} \lambda_j \left| \frac{1}{n} \sum_{i=1}^n \sum_{l=1}^n w_{l,0}^{(i)} [\langle Z_l, \hat{v}_j - v_j \rangle (\langle Z_l, \boldsymbol{\theta}_0 \rangle + \epsilon_l)] \right|^2 / \hat{\lambda}_j\right) \\
& \leq 2\left(\sup_{j \leq k_n} \frac{\lambda_j}{\lambda_{j+1}}\right) \times \left(\sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n E|\langle Z_i, \hat{v}_j - v_j \rangle (\langle Z_i, \boldsymbol{\theta}_0 \rangle + \epsilon_i)|^2\right) \\
& + 2\left(\sup_{j \leq k_n} \frac{\lambda_j}{\lambda_{j+1}}\right) \times \left(\sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n E\left|\sum_{l=1}^n w_{l,0}^{(i)} [\langle Z_l, \hat{v}_j - v_j \rangle (\langle Z_l, \boldsymbol{\theta}_0 \rangle + \epsilon_l)]\right|^2\right) \\
& \leq 2\left(\sup_{j \leq k_n} \frac{\lambda_j}{\lambda_{j+1}}\right) \times \left(\sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n E|\langle Z_i, \hat{v}_j - v_j \rangle (\langle Z_i, \boldsymbol{\theta}_0 \rangle + \epsilon_i)|^2\right) \\
& + 2\left(\sup_{j \leq k_n} \frac{\lambda_j}{\lambda_{j+1}}\right) \times \left(\sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n E \sum_{l=1}^n (w_{l,0}^{(i)})^2 \sum_{l=1}^n [\langle Z_l, \hat{v}_j - v_j \rangle (\langle Z_l, \boldsymbol{\theta}_0 \rangle + \epsilon_l)]^2\right) \\
& \leq 4\left(\sup_{j \leq k_n} \frac{\lambda_j}{\lambda_{j+1}}\right) \times \left(\sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n E|\langle Z_i, \hat{v}_j - v_j \rangle (\langle Z_i, \boldsymbol{\theta}_0 \rangle + \epsilon_i)|^2\right) \\
& \leq \text{constant} \times k_n^{a-1} \sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n E[\langle Z_i, Z_i \rangle \|\hat{v}_j - v_j\|^2 (\langle Z_i, \boldsymbol{\theta}_0 \rangle + \epsilon_i)^2] \\
& \leq \text{constant} \times k_n^{a-1} \sum_{j=1}^{k_n} \zeta_j^{-2} n^{-1} = O(k_n^{a-1} \sum_{j=1}^{k_n} j^{3a+2}/n) = O(k_n^{4a+2} n^{-1}).
\end{aligned}$$

□

Proof of Lemma 12 (iii). After straightforward algebraic calculation, $\sum_{j=1}^{k_n} \lambda_j^2 (\hat{r}_j^* - r_j^*)^2 =$

$$\begin{aligned}
& \sum_{j=1}^{k_n} \left[\frac{1}{n} \sum_{i=1}^n (\hat{m}_{M_j, i, \gamma_0}(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_{1,0}) - m_{M_j, i, \gamma_0}^0(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_{1,0})) \right]^2 \\
& \leq \sum_{j=1}^{k_n} \frac{1}{n} \sum_{i=1}^n (\hat{m}_{M_j, i, \gamma_0}(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1) - m_{M_j, i, \gamma_0}^0(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_{1,0}))^2.
\end{aligned}$$

Next we calculate $|\hat{m}_{M_j, i, \gamma_0}(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1) - m_{M_j, i, \gamma_0}^0(Y_i, Z_i, W_i, v_j; \boldsymbol{\beta}_1)|$. Denote

$$\begin{aligned}
A_i^{(12,3)} &= 1/n \sum_{l=1}^n K_h^{(l)}(w_0) \sqrt{\sum_{j_1=1}^{k_n} \langle Z_l - Z_i, v_{j_1} \rangle^2 + (1 - w_0) \|W_l - W_i\|}, \\
B_i^{(12,3)} &= 1/n \sum_{l=1}^n \tilde{K}_h^{(l)}(w_0) \sqrt{\sum_{j_1=1}^{k_n} \langle Z_l - Z_i, v_{j_1} \rangle^2 + (1 - w_0) \|W_l - W_i\|}, \\
C_i^{(12,3)} &= E(\delta_i \exp(\gamma_0 Y_i) \langle \boldsymbol{\theta}_0, Z_i \rangle | Z_i, W_i), D_i^{(12,3)} = E(\delta_i \exp(\gamma_0 Y_i) | Z_i, W_i),
\end{aligned}$$

and

$$E_i^{(12,3)} = E[K_h(w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z - Z_i, v_{j_1} \rangle^2 + (1-w_0)\|W - W_i\|}) | Z_i, W_i].$$

By Condition (B.1), we have $D_i^{(12,3)} \geq$

$$\begin{aligned} & \inf_{\max\{\|z\|, \|x\|\} \leq C_1} E(\delta_i | Z_i = z, W_i = x) \\ &= \inf_{\max\{\|z\|, \|x\|\} \leq C_1} \int \frac{\phi y + G(z, x)}{1 + \exp(\phi_0 Y_i + G(z, x))} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(y - \beta_{1,0}^T x - \langle \theta_0, z \rangle)^2} dy \\ &\geq \text{constant} > 0. \end{aligned}$$

The last inequality is because the integral is a continuous function of $(x, G(z, x), \langle z, \theta_0 \rangle)$ and each item of $x, G(z, x), \langle z, \theta_0 \rangle$ is positive in a compact set from Condition (B.1) and (B.2). From $|\hat{m}_{M_j, i, \gamma_0}(Y_i, Z_i, W_i, v_j; \beta_1) - m_{M_j, i, \gamma_0}^0(Y_i, Z_i, W_i, v_j; \beta_1)| =$

$$\langle Z_i, v_j \rangle \left| \frac{A_i^{(12,3)}}{B_i^{(12,3)}} - \frac{C_i^{(12,3)}}{D_i^{(12,3)}} \right| + \frac{1/n \sum_{l=1}^n K_h^{(l)}(w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z_l - Z_i, v_{j_1} \rangle^2 + (1-w_0)\|W_l - W_i\|}) \langle Z_l - Z_i, v_j \rangle}{B_i^{(12,3)}},$$

we have $\sum_{j=1}^{\infty} \sum_{i=1}^n |\hat{m}_{M_j, i, \gamma_0}(Y_i, Z_i, W_i, v_j; \beta_1) - m_{M_j, i, \gamma_0}^0(Y_i, Z_i, W_i, v_j; \beta_1)|^2/n \leq F_1^{(12,3)} + F_2^{(12,3)}$, where

$$\begin{aligned} F_1^{(12,3)} &= \sum_{j=1}^{\infty} \frac{2}{n} \sum_{i=1}^n \langle Z_i, v_j \rangle \left| \frac{A_i^{(12,3)}}{B_i^{(12,3)}} - \frac{C_i^{(12,3)}}{D_i^{(12,3)}} \right|^2; \\ F_2^{(12,3)} &= \sum_{j=1}^{\infty} \frac{2}{n} \sum_{i=1}^n \left[\frac{1/n \sum_{l=1}^n K_h^{(l)}(w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z_l - Z_i, v_{j_1} \rangle^2 + (1-w_0)\|W_l - W_i\|}) \langle Z_l - Z_i, v_j \rangle}{B_i^{(12,3)}} \right]^2, \end{aligned}$$

Next we calculate the two terms $F_1^{(12,3)}$ and $F_2^{(12,3)}$ separately. By Lemma 11, (ii), the following equations hold uniformly with respect to (z, x) .

$$\frac{A_i^{(12,3)}}{E_i^{(12,3)}} - C_i^{(12,3)} \leq \text{constant} \times \sqrt{\left(h^2 + \frac{1}{n\psi(h)}\right)}; \quad \frac{B_i^{(12,3)}}{E_i^{(12,3)}} - D_i \leq \text{constant} \times \sqrt{h^2 + \frac{1}{n\psi(h)}}.$$

It follows that $\sup_{z,x} |A_i^{(12,3)}/B_i^{(12,3)} - C_i^{(12,3)}/D_i^{(12,3)}| =$

$$\begin{aligned} & \left| \frac{A_i^{(12,3)}/E_i^{(12,3)} - C_i^{(12,3)}}{D_i^{(12,3)} + (B_i^{(12,3)}/E_i^{(12,3)} - D_i^{(12,3)})} - \frac{C_i^{(12,3)}(D_i^{(12,3)} - B_i^{(12,3)}/E_i^{(12,3)})}{[D_i^{(12,3)} + (B_i^{(12,3)}/E_i^{(12,3)} - D_i^{(12,3)})]D_i^{(12,3)}} \right| \\ &\leq \text{constant} \times \sqrt{\left(h^2 + \frac{1}{n\psi(h)}\right)}. \end{aligned}$$

Then we have

$$F_1^{(12,3)} \leq \text{constant} \times \left(h^2 + \frac{1}{n\psi(h)}\right) \sum_{j=1}^{\infty} \frac{1}{n} \sum_{i=1}^n \langle Z_i, v_j \rangle^2 = O_p\left(h^2 + \frac{1}{n\psi(h)}\right).$$

Note that $F_2^{(12,3)}$ can be simplified as

$$\frac{2}{n} \sum_{i=1}^n \left[\frac{1/n \sum_{l=1}^n K_h^{(l)}(w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z_l - Z_i, v_{j_1} \rangle^2 + (1-w_0)\|W_l - W_i\|}) \|Z_l - Z_i\|^2}{B_i^{(12,3)}} \right]^2$$

which equals (similar to Lemma 11 (ii))

$$\begin{aligned} & \frac{2}{n} \sum_{i=1}^n \left[\frac{1/n \sum_{l=1}^n K_h^{(l)}(w_0 \sqrt{\sum_{j=1}^{k_n} \langle Z_l - Z_i, v_{j_1} \rangle^2} + (1-w_0) \|W_l - W_i\|) 1_{\|Z_l - Z_i\| \leq \frac{h}{w_0}} \|Z_l - Z_i\|^2}{B_i^{(12,3)}} \right]^2 \\ &= O_p(h^2). \end{aligned}$$

Finally, we have

$$\sum_{j=1}^{k_n} \lambda_j^2 (\hat{r}_j^* - r_j^*)^2 = F_1^{(12,3)} + F_2^{(12,3)} = O_p(h^2 + \frac{1}{n\psi(h)}).$$

□

Proof of Lemma 12 (iv). Denote

$$\Delta_1 \left[\sum_{j=1}^{k_n} \hat{r}_j (\delta_i W_i \langle Z_i, \hat{v}_j \rangle) \right] = \sum_{j=1}^{k_n} \hat{r}_j (\delta_i W_i \langle Z_i, \hat{v}_j \rangle) - \sum_{j=1}^{k_n} \hat{r}_j^* (\delta_i W_i \langle Z_i, v_j \rangle),$$

and

$$\Delta_1 \left[\sum_{j=1}^{k_n} \hat{r}_j \left(\sum_{l=1}^n w_{l,0} \delta_l W_l \langle Z_l, \hat{v}_j \rangle \right) \right] = \sum_{j=1}^{k_n} \hat{r}_j \left(\sum_{l=1}^n w_{l,0} \delta_l W_l \langle Z_l, \hat{v}_j \rangle \right) - \sum_{j=1}^{k_n} \hat{r}_j^* \left(\sum_{l=1}^n w_{l,0} \delta_l W_l \langle Z_l, v_j \rangle \right).$$

We have the decomposition similar to the proof of Lemma 6, (vi).

$$\begin{aligned} & \frac{1}{n} \sum_{i=1}^n \Delta_1 \left[\sum_{j=1}^{k_n} \hat{r}_j (\delta_i W_i \langle Z_i, \hat{v}_j \rangle) \right] \\ &= \sum_{j=1}^{k_n} (\hat{r}_j - \hat{r}_j^*) \frac{1}{n} \sum_{i=1}^n \delta_i W_i \langle Z_i, \hat{v}_j - v_j \rangle + \sum_{j=1}^{k_n} (\hat{r}_j - \hat{r}_j^*) \frac{1}{n} \sum_{i=1}^n \delta_i W_i \langle Z_i, v_j \rangle \\ &+ \sum_{j=1}^{k_n} (\hat{r}_j^* - r_j^*) \frac{1}{n} \sum_{i=1}^n \delta_i W_i \langle Z_i, \hat{v}_j - v_j \rangle + \sum_{j=1}^{k_n} (r_j^* - \langle \theta_0, v_j \rangle) \frac{1}{n} \sum_{i=1}^n \delta_i W_i \langle Z_i, \hat{v}_j - v_j \rangle \\ &+ \sum_{j=1}^{k_n} \langle \theta_0, v_j \rangle \frac{1}{n} \sum_{i=1}^n \delta_i W_i \langle Z_i, \hat{v}_j - v_j \rangle \triangleq A^{(12,4)} + B^{(12,4)} + C^{(12,4)} + D^{(12,4)} + E^{(12,4)}. \end{aligned}$$

Similar to the proof of Lemma 6, (vi), and using Lemma 12, (i), and Lemma 12, (ii), we have $A^{(12,4)} + B^{(12,4)} + D^{(12,4)} + E^{(12,4)} = O_p(k_n^{2a+1} n^{-1/2} + k_n^{-b})$. We only need to calculate the order of the term $C^{(12,4)}$. Using the Cauchy's inequality and Lemma 12, (iii), we have

$$\begin{aligned} C^{(12,4)} &\leq \sqrt{\sum_{j=1}^{k_n} \lambda_j^2 (\hat{r}_j^* - r_j^*)^2 \sum_{j=1}^{k_n} \left[\left(\frac{1}{n} \sum_{i=1}^n \delta_i W_i \langle Z_i, \hat{v}_j - v_j \rangle \right) / \lambda_j \right]^2} \\ &= \sqrt{O_p \left[h^2 + \frac{1}{n\psi(h)} \sum_{j=1}^{k_n} 1/\zeta_j^2 / \lambda_j^2 \times 1/n \right]} = O_p \left(\frac{k_n^{(4a+3)/2} \left[h + \frac{1}{\sqrt{n\psi(h)}} \right]}{\sqrt{n}} \right). \end{aligned}$$

Then we have $\frac{1}{n} \sum_{i=1}^n \Delta_1 \left[\sum_{j=1}^{k_n} \hat{r}_j (\delta_i W_i \langle Z_i, \hat{v}_j \rangle) \right] =$

$$A^{(12,4)} + B^{(12,4)} + C^{(12,4)} + D^{(12,4)} + E^{(12,4)} = O_p(k_n^{2a+1} n^{-1/2} + k_n^{-b} + \frac{k_n^{(4a+3)/2}}{\sqrt{n}} \left[h + \frac{1}{\sqrt{n\psi(h)}} \right]).$$

Similarly, we can also prove

$$\begin{aligned} & \frac{1}{n} \sum_{i=1}^n \Delta_1 \left[\sum_{j=1}^{k_n} \hat{r}_j \left(\sum_{l=1}^n w_{l,0} \delta_l W_l \langle Z_l, \hat{v}_j \rangle \right) \right] \\ &= O_p(k_n^{2a+1} n^{-1/2} + k_n^{-b} + k_n^{(4a+3)/2} n^{-1/2} [h + \frac{1}{\sqrt{n\psi(h)}}]). \end{aligned}$$

The conclusion holds based on the above results. \square

Lemma 13. *Under Assumption A.1–A.9 and B.1–B.5, we have*

$$\begin{aligned} L_{13} &\triangleq \frac{1}{n} \sum_{i=1}^n \left\{ \sum_{j=1}^{k_n} \hat{r}_j^* [\delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) \sum_{l=0}^n w_{l,0} (Z_i, W_i; \gamma) \langle Z_l, v_j \rangle W_l] \right\} \\ &\quad - \frac{1}{n} \sum_{i=1}^n \left\{ \sum_{j=1}^{k_n} r_j^* [\delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) E(\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0)] \right\} \\ &= O_p(k_n^{(a+1)/2} [h + 1/\sqrt{n\psi(h)}]). \end{aligned}$$

Proof of Lemma 13. First we decompose L_{13} in the following.

$$\begin{aligned} L_{13} &\leq \frac{1}{n} \sum_{i=1}^n \left\{ \sum_{j=1}^{k_n} |\hat{r}_j^* - r_j^*| \times |E(\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0) - \sum_{l=0}^n w_{l,0} (Z_i, W_i; \gamma) \langle Z_l, v_j \rangle W_l| \right\} \\ &\quad + \frac{1}{n} \sum_{i=1}^n \left\{ \sum_{j=1}^{k_n} |\hat{r}_j^* - r_j^*| \times [\delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) E(\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0)] \right\} \\ &\quad + \frac{1}{n} \sum_{i=1}^n \left\{ \sum_{j=1}^{k_n} |r_j^* - \langle \theta_0, v_j \rangle| \times |E(\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0) - \sum_{l=0}^n w_{l,0} (Z_i, W_i; \gamma) \langle Z_l, v_j \rangle W_l| \right\} \\ &\quad + \frac{1}{n} \sum_{i=1}^n \left\{ \sum_{j=1}^{k_n} |\langle \theta_0, v_j \rangle| \times |E(\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0) - \sum_{l=0}^n w_{l,0} (Z_i, W_i; \gamma) \langle Z_l, v_j \rangle W_l| \right\} \\ &\triangleq A^{(13)} + B^{(13)} + C^{(13)} + D^{(13)}. \end{aligned}$$

Define $\widehat{K}_h^{(l)}(\cdot) = K_h(\cdot) \delta_l \exp(\gamma_0 Y_l) W_l$. Similar to the proof of Lemma 11, (ii), we have

$$\begin{aligned} & \sup_{z,x} E \left\{ \frac{1/n \sum_{l=1}^n \widehat{K}_h^{(l)}(w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z_l - Z_i, v_{j_1} \rangle^2} + (1 - w_0) \|W_l - W_i\|)}{E[K_h(w_0 \sqrt{\sum_{j_1=1}^{k_n} \langle Z - Z_i, v_{j_1} \rangle^2} + (1 - w_0) \|W - W_i\|) | Z_i, W_i]} \right. \\ & \quad \left. - E(W_i \delta_i \exp(\gamma Y_i) | Z_i, W_i) \right]^2 | Z_i = z, W_i = x \} \\ & \leq \text{constant} \times (h^2 + \frac{1}{n\psi(h)}). \end{aligned}$$

Consequently, similar to the proof of Lemma 12, (iii), we get

$$\sum_{j=1}^{k_n} |E(\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0) - \sum_{l=0}^n w_{l,0} (Z_i, W_i; \gamma) \langle Z_l, v_j \rangle W_l|^2 = O_p(h^2 + \frac{1}{n\psi(h)}).$$

Then from Lemma 6, we have

$$\begin{aligned}
C^{(13)} &\leq \sqrt{\sum_{j=1}^{k_n} (r_j^* - \langle \boldsymbol{\theta}_0, v_j \rangle)^2} \\
&\times \sqrt{\sum_{j=1}^{k_n} |\mathbb{E}(\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0) - \sum_{l=0}^n w_{l,0}(Z_i, W_i; \gamma) \langle Z_l, v_j \rangle W_l|^2} \\
&= \sqrt{O_p(k_n^{a+1}/n) O_p(h^2 + \frac{1}{n\psi(h)})} = O_p(k_n^{(a+1)/2} n^{-1/2} [h + \frac{1}{\sqrt{n\psi(h)}}]),
\end{aligned}$$

and

$$\begin{aligned}
D^{(13)} &\leq \sqrt{\sum_{j=1}^{k_n} \langle \boldsymbol{\theta}_0, v_j \rangle^2} \\
&\times \sqrt{\sum_{j=1}^{k_n} |\mathbb{E}(\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0) - \sum_{l=0}^n w_{l,0}(Z_i, W_i; \gamma) \langle Z_l, v_j \rangle W_l|^2} \\
&= \sqrt{O_p(1) O_p(h^2 + \frac{1}{n\psi(h)})} = O_p(h + \frac{1}{\sqrt{n\psi(h)}}).
\end{aligned}$$

Similarly, using the Cauchy's inequality and Lemma 12, (iii), we have

$$\begin{aligned}
A^{(13)} &\leq \sqrt{\sum_{j=1}^{k_n} (\hat{r}_j^* - r_j^*)^2} \\
&\times \sqrt{\sum_{j=1}^{k_n} |\mathbb{E}(\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0) - \sum_{l=0}^n w_{l,0}(Z_i, W_i; \gamma) \langle Z_l, v_j \rangle W_l|^2} \\
&= \sqrt{O_p(k_n^{2a} [h^2 + \frac{1}{n\psi(h)}]) O_p(h^2 + \frac{1}{n\psi(h)})} = O_p(k_n^a [h^2 + \frac{1}{n\psi(h)}]).
\end{aligned}$$

Similarly, using the Cauchy's inequality and Lemma 6, (iv), we have

$$\begin{aligned}
B^{(13)} &\leq \sqrt{\sum_{j=1}^{k_n} \lambda_j^2 (\hat{r}_j^* - r_j^*)^2} \\
&\times \sqrt{\sum_{j=1}^{k_n} \{ \frac{1}{n} \sum_{i=1}^n [\delta_i W_i \langle Z_i, v_j \rangle + (1 - \delta_i) \mathbb{E}(\langle Z_i, v_j \rangle W | Z_i, W_i, \delta_i = 0)]^2 / \lambda_j^2} \\
&= \sqrt{O_p(h^2 + \frac{1}{n\psi(h)}) O_p(k_n^{a+1})} = O_p(k_n^{(a+1)/2} [h^2 + \frac{1}{n\psi(h)}]).
\end{aligned}$$

Finally, we get $L_{13} = O_p(A^{(13)} + B^{(13)} + C^{(13)} + D^{(13)}) = O_p(k_n^{(a+1)/2} [h + 1/\sqrt{n\psi(h)}])$.

□

Proof of Theorem 2.

From Lemma 4, 5, 7, 12 and 13, we get

$$\begin{aligned}\tilde{U}(\boldsymbol{\beta}_{1,0}) &= L_5 - L_4 + L_7 + L_{12} + L_{13} \\ &= O_p(k_n^{1/2-b} + k_n n^{-1/2} + k_n^{2a+1} n^{-1/2} + k_n^{-b} + k_n^{(4a+3)/2} n^{-1/2} [h + \frac{1}{\sqrt{n\psi(h)}}] + k_n^{(a+1)/2} [h + 1/\sqrt{n\psi(h)}]) \\ &= O_p(k_n^{1/2-b} + k_n^{2a+1} n^{-1/2} + k_n^{(a+1)/2} [h + 1/\sqrt{n\psi(h)}]),\end{aligned}$$

where the ' L_i ' is defined in Lemma i , $i = 4, 5, 7, 12, 13$. Similar to Lemma 9, we have $\|\partial^2 \tilde{U}(\boldsymbol{\beta}_{1,0})/(\partial \boldsymbol{\beta}_1 \partial \boldsymbol{\beta}_1^T) - \mathfrak{J}\| = o(1)$. Then we have

$$\begin{aligned}\|\tilde{\boldsymbol{\beta}}_1 - \boldsymbol{\beta}_{1,0}\| &= \|\mathfrak{J}^{-1} \tilde{U}(\boldsymbol{\beta}_{1,0})\| (1 + o_p(1)) \\ &= O_p(k_n^{1/2-b} + k_n^{2a+1} n^{-1/2} + k_n^{(a+1)/2} [h + 1/\sqrt{n\psi(h)}]),\end{aligned}$$

and the first part of Theorem 2 is finished.

We continue with the second part of the theorem. For $\tilde{\boldsymbol{\theta}} = \sum_{j=1}^{k_n} r_j(\tilde{\boldsymbol{\beta}}_1) \hat{v}_j$, we have

$$\begin{aligned}\tilde{\boldsymbol{\theta}} - \boldsymbol{\theta}_0 &= \left\{ \sum_{j=1}^{k_n} (r_j(\tilde{\boldsymbol{\beta}}_1) - r_j(\boldsymbol{\beta}_{1,0})) \hat{v}_j \right\} + \left\{ \sum_{j=1}^{k_n} r_j(\boldsymbol{\beta}_{1,0}) \hat{v}_j - \sum_{j=1}^{k_n} \langle \boldsymbol{\theta}_0, v_j \rangle v_j \right\} \\ &+ \left\{ - \sum_{j=k_n+1}^{\infty} \langle \boldsymbol{\theta}_0, v_j \rangle v_j \right\} + \sum_{j=1}^{k_n} (\hat{r}_j(\tilde{\boldsymbol{\beta}}_1) - r_j(\tilde{\boldsymbol{\beta}}_1)) \hat{v}_j \\ &\triangleq A^{\mathbf{T}2} + B^{\mathbf{T}2} + C^{\mathbf{T}2} + D^{\mathbf{T}2}.\end{aligned}$$

By the proof of Theorem 1, $\|C^{\mathbf{T}2}\| = \|C^{\mathbf{T}1}\| = O_p(k_n^{1/2-b})$; $\|B^{\mathbf{T}2}\| = \|B^{\mathbf{T}1}\| = O_p(k_n^{5a/2+3/2}/\sqrt{n} + k_n^{a/2+1/2-b})$; and

$$\begin{aligned}\|A^{\mathbf{T}2}\| &\leq \|A_1^{\mathbf{T}2} + A_2^{\mathbf{T}2} + A_3^{\mathbf{T}2} + A_4^{\mathbf{T}2}\| \|\tilde{\boldsymbol{\beta}}_1 - \boldsymbol{\beta}_{1,0}\| \\ &= O_p(k_n^{(a+1)/2}) O_p(k_n^{2a+1} n^{-1/2} + k_n^{1/2-b} + k_n^{(a+1)/2} [h + 1/\sqrt{n\psi(h)}]) \\ &= O_p(k_n^{5/2a+3/2} n^{-1/2} + k_n^{1+a/2-b} + k_n^{a+1} [h + 1/\sqrt{n\psi(h)}]),\end{aligned}$$

where $A_1^{\mathbf{T}2} = A_1^{\mathbf{T}1}$, $A_2^{\mathbf{T}2} = A_1^{\mathbf{T}1}$, $A_3^{\mathbf{T}2} = A_1^{\mathbf{T}1}$ and $A_4^{\mathbf{T}2} = A_1^{\mathbf{T}1}$, and $A_i^{\mathbf{T}1}$ are defined in Theorem 1, for $i = 1, 2, 3, 4$. So that we only need to calculate $\|D^{\mathbf{T}2}\|$. First we define

$$\tilde{r}_j = \frac{\sum_{i=1}^n [\delta_i \tilde{M}_j(Z_i, W_i, \hat{v}_j) + (1 - \delta_i) \hat{m}_{\tilde{M}_j, i, \gamma}(Z_i, W_i, \hat{v}_j)]}{n \hat{\lambda}_j},$$

and

$$\tilde{r}_j^* = \frac{\sum_{i=1}^n [\delta_i \tilde{M}_j(Z_i, W_i, v_j) + (1 - \delta_i) \hat{m}_{\tilde{M}_j, i, \gamma}(Z_i, W_i, v_j)]}{n \lambda_j},$$

where the definition of $\tilde{M}_j(Z_i, W_i, \hat{v}_j)$ can be found in the proof of Lemma 9. Then similar to the proof of Theorem 1, we have

$$\begin{aligned}\|D^{\mathbf{T}2}\| &= \sqrt{\sum_{j=1}^{k_n} [\hat{r}_j(\tilde{\boldsymbol{\beta}}_1) - r_j(\tilde{\boldsymbol{\beta}}_1)]^2} = \sqrt{\sum_{j=1}^{k_n} [\hat{r}_j(\boldsymbol{\beta}_{1,0}) - r_j(\boldsymbol{\beta}_{1,0}) + (\hat{r}_j - \tilde{r}_j)(\tilde{\boldsymbol{\beta}}_1 - \boldsymbol{\beta}_{1,0})]^2} \\ &\leq \sqrt{\sum_{j=1}^{k_n} [\hat{r}_j(\boldsymbol{\beta}_{1,0}) - r_j(\boldsymbol{\beta}_{1,0})]^2} + \sqrt{\sum_{j=1}^{k_n} [(\hat{r}_j - \tilde{r}_j)(\tilde{\boldsymbol{\beta}}_1 - \boldsymbol{\beta}_{1,0})]^2} \triangleq D_1^{\mathbf{T}2} + D_2^{\mathbf{T}2},\end{aligned}$$

By the conclusions of Lemma 12, (ii), Lemma 12, (iii), Lemma 6, (iii), and the Cauchy's inequality, we have

$$\begin{aligned}
D_1^{\mathbf{T}2} &= \sqrt{\sum_{j=1}^{k_n} [(\hat{r}_j - \hat{r}_j^*) + (\hat{r}_j^* - r_j^*) + (r_j^* - r_j)]^2} \\
&\leq \sqrt{\sum_{j=1}^{k_n} (\hat{r}_j - \hat{r}_j^*)^2} + \sqrt{\sum_{j=1}^{k_n} (\hat{r}_j^* - r_j^*)^2} + \sqrt{\sum_{j=1}^{k_n} (r_j^* - r_j)^2} \\
&\leq \sqrt{\sum_{j=1}^{k_n} \lambda_j (\hat{r}_j - \hat{r}_j^*)^2} \sum_{j=1}^{k_n} \frac{1}{\lambda_j} + \sqrt{\sum_{j=1}^{k_n} \lambda_j^2 (\hat{r}_j^* - r_j^*)^2} \sum_{j=1}^{k_n} \frac{1}{\lambda_j^2} + \sqrt{\sum_{j=1}^{k_n} \lambda_j (r_j^* - r_j)^2} \sum_{j=1}^{k_n} \frac{1}{\lambda_j} \\
&= \sqrt{O_p(k_n^{4a+2} n^{-1} + k_n^{-2b}) \times k_n^{a+1}} + \sqrt{O_p(h^2 + \frac{1}{n\psi(h)}) \times k_n^{2a+1}} \\
&\quad + \sqrt{O_p(k_n^{4a+2} n^{-1} + k_n^{-2b}) \times k_n^{a+1}} \\
&= O_p(k_n^{5a/2+3/2} n^{-1/2} + k_n^{1+a/2-b} + k_n^{a+1/2} [h + \frac{1}{\sqrt{n\psi(h)}}]).
\end{aligned}$$

Similar to the calculation of $D_1^{\mathbf{T}2}$, we have

$$\sqrt{\sum_{j=1}^{k_n} [(\tilde{r}_j - \tilde{r}_j)^2]} = O_p(k_n^{5a/2+3/2} n^{-1/2} + k_n^{1+a/2-b} + k_n^{a+1/2} [h + \frac{1}{\sqrt{n\psi(h)}}]).$$

Note that $D_2^{\mathbf{T}2} = \sqrt{\sum_{j=1}^{k_n} [(\tilde{r}_j - \tilde{r}_j)^2]} \|\tilde{\beta}_1 - \beta_{1,0}\|$, which implies $D_2^{\mathbf{T}2} = o_p(D_1^{\mathbf{T}2})$.

Finally, we get the conclusion.

$$\begin{aligned}
\|\tilde{\theta} - \theta_0\| &\leq \|A^{\mathbf{T}2}\| + \|B^{\mathbf{T}2}\| + \|C^{\mathbf{T}2}\| + \|D^{\mathbf{T}2}\| \\
&= O_p(k_n^{5a/2+3/2} n^{-1/2} + k_n^{1+a/2-b} + k_n^{a+1} [h + \frac{1}{\sqrt{n\psi(h)}}]).
\end{aligned}$$

□

Proof of Theorem 3

First of all, minimizing the following expression

$$(\mathbf{Y} - Z^* \mathbf{r} - \mathbf{W}\beta)^T \Sigma (\mathbf{Y} - Z^* \mathbf{r} - \mathbf{W}\beta) + (Z^* \mathbf{r})^T (I_n - \Sigma) Z^* \mathbf{r},$$

where $\Sigma = \{D + \text{diag}[\Xi(I_n - D)\mathbf{1}_n]\}$, is equivalent to solving the following eqnarray

$$\begin{cases} \frac{1}{n} Z^{*T} \{D + \text{diag}[\Xi(I_n - D)\mathbf{1}_n]\} (\mathbf{Y} - \mathbf{W}\beta) - \frac{1}{n} Z^{*T} Z^* \mathbf{r} = 0, \\ \frac{1}{n} \mathbf{W}^T \{D + \text{diag}[\Xi(I_n - D)\mathbf{1}_n]\} (\mathbf{Y} - Z^* \mathbf{r} - \mathbf{W}\beta) = 0. \end{cases} \quad (\text{S1.4})$$

From the definitions of Subsection 2.2.2 of the main text, $Z^* = \begin{pmatrix} \bar{Z}_1 \bar{V}_{k_n} \\ \vdots \\ \bar{Z}_n \bar{V}_{k_n} \end{pmatrix}$, where $\bar{Z}_i \triangleq$

$(\bar{Z}_{i,1}, \bar{Z}_{i,2}, \dots, \bar{Z}_{i,n}), i = 1, 2, \dots, n$, and

$$\Xi(I_n - D)\mathbf{1}_n = \begin{pmatrix} \sum_{i=1}^n (1 - \delta_i) w_{1,i} \\ \sum_{i=1}^n (1 - \delta_i) w_{2,i} \\ \vdots \\ \sum_{i=1}^n (1 - \delta_i) w_{n,i} \end{pmatrix}.$$

Then we have the following equivalence.

$$\begin{aligned} & \frac{1}{n} Z^{*T} \{D + \text{diag}[\Xi(I_n - D)\mathbf{1}_n]\} (\mathbf{Y} - \mathbf{W}\boldsymbol{\beta}) - \frac{1}{n} Z^{*T} Z^* \mathbf{r} = 0 \\ \iff & \frac{1}{n} \begin{pmatrix} \bar{Z}_1 \bar{V}_{k_n} \\ \vdots \\ \bar{Z}_n \bar{V}_{k_n} \end{pmatrix}^T \text{diag} \left\{ \delta_1 + \sum_{i=1}^n (1 - \delta_i) w_{1,i}, \dots, \delta_n + \sum_{i=1}^n (1 - \delta_i) w_{n,i} \right\} \\ & \times \begin{pmatrix} Y_1 - W_1 \boldsymbol{\beta} \\ \vdots \\ Y_n - W_n \boldsymbol{\beta} \end{pmatrix} - \hat{\Lambda} \mathbf{r} = 0 \\ \iff & \frac{1}{n} \sum_{i=1}^n (\bar{Z}_i \bar{V}_{k_n})^T [\delta_i + \sum_{k=1}^n (1 - \delta_k) w_{i,k}] (Y_i - W_i \boldsymbol{\beta}) - \sum_{i=1}^n [\delta_i + \sum_{k=1}^n (1 - \delta_k) w_{i,k}] \frac{\hat{\Lambda}}{n} \mathbf{r} = 0 \\ \iff & \frac{1}{n} \sum_{i=1}^n (\bar{Z}_i \bar{V}_{k_n})^T \delta_i (Y_i - W_i \boldsymbol{\beta}) + \sum_{i,k \leq n} (\bar{Z}_i \bar{V}_{k_n})^T (1 - \delta_k) w_{i,k} (Y_i - W_i \boldsymbol{\beta}) \\ & - \sum_{i=1}^n [\delta_i + \sum_{k=1}^n (1 - \delta_k) w_{i,k}] \frac{\hat{\Lambda}}{n} \mathbf{r} = 0 \\ \iff & \frac{1}{n} \sum_{i=1}^n (\bar{Z}_i \bar{V}_{k_n})^T \delta_i (Y_i - W_i \boldsymbol{\beta}) + \sum_{i,l \leq n} (\bar{Z}_i \bar{V}_{k_n})^T (1 - \delta_l) w_{l,i} (Y_l - W_l \boldsymbol{\beta}) \\ & - \sum_{i=1}^n [\delta_i + \sum_{k=1}^n (1 - \delta_k) w_{i,k}] \frac{\hat{\Lambda}}{n} \mathbf{r} = 0 \\ \iff & \frac{1}{n} \sum_{i=1}^n [\delta_i (\bar{Z}_i \bar{V}_{k_n})^T (Y_i - W_i \boldsymbol{\beta}) + (1 - \delta_i) \sum_{l=1}^n w_{l,i} (\bar{Z}_l \bar{V}_{k_n})^T (Y_l - W_l \boldsymbol{\beta})] \\ & - \sum_{i=1}^n [\delta_i + \sum_{l=1}^n (1 - \delta_l) w_{l,i}] \frac{\hat{\Lambda}}{n} \mathbf{r} = 0 \\ \iff & \frac{1}{n} \sum_{i=1}^n \left\{ \delta_i [(\bar{Z}_i \bar{V}_{k_n})^T (Y_i - W_i \boldsymbol{\beta}) - \begin{pmatrix} \hat{\lambda}_1 \mathbf{r}_1 \\ \vdots \\ \hat{\lambda}_{k_n} \mathbf{r}_{k_n} \end{pmatrix}] + (1 - \delta_i) \times \right. \\ & \left. \sum_{l=1}^n w_{l,i} [(\bar{Z}_l \bar{V}_{k_n})^T (Y_l - W_l \boldsymbol{\beta}) - \begin{pmatrix} \hat{\lambda}_1 \mathbf{r}_1 \\ \vdots \\ \hat{\lambda}_{k_n} \mathbf{r}_{k_n} \end{pmatrix}] \right\} = 0 \end{aligned}$$

$$\begin{aligned} &\iff \\ &\sum_{i=1}^n [\delta_i \psi_1(Y_i, Z_i, W_i; \mathbf{r}, \boldsymbol{\beta}_1) + (1 - \delta_i) \hat{m}_{\psi_1, i, \gamma}(Y_i, Z_i, W_i; \mathbf{r}, \boldsymbol{\beta}_1)] = 0, \end{aligned}$$

where

$$\psi_1(Y_i, Z_i, W_i; \mathbf{r}, \boldsymbol{\beta}_1) = (\bar{Z}_i \bar{V}_{k_n})^T (Y_i - W_i \boldsymbol{\beta}) - \begin{pmatrix} \hat{\lambda}_1 \mathbf{r}_1 \\ \vdots \\ \hat{\lambda}_{k_n} \mathbf{r}_{k_n} \end{pmatrix},$$

is a discretized form of

$$\begin{pmatrix} \langle Z_i, \hat{v}_1 \rangle (Y_i - W_i \boldsymbol{\beta}) - \hat{\lambda}_1 r_1 \\ \vdots \\ \langle Z_i, \hat{v}_{k_n} \rangle (Y_i - W_i \boldsymbol{\beta}) - \hat{\lambda}_{k_n} r_{k_n} \end{pmatrix}. \quad (\text{S1.5})$$

Similarly, the second equation of (??) is equivalent to

$$\sum_{i=1}^n [\delta_i \psi_2(Y_i, Z_i, W_i; \mathbf{r}, \boldsymbol{\beta}_1) + (1 - \delta_i) \hat{m}_{\psi_2, i, \gamma}(Y_i, Z_i, W_i; \mathbf{r}, \boldsymbol{\beta}_1)] = 0,$$

where $\psi_2(Y_i, Z_i, W_i; \mathbf{r}, \boldsymbol{\beta}_1) = [Y_i - \boldsymbol{\beta}_1^T W_i - \bar{Z}_i \bar{V}_{k_n} \mathbf{r}] W_i$ is a discretized form of

$$W_i^T [Y_i - \boldsymbol{\beta}_1^T W_i - \sum_{j=1}^{k_n} r_j \langle Z_i, v_j \rangle]. \quad (\text{S1.6})$$

Compare (??) and (??) with (2.5), and we get the conclusion in Theorem 3.

□

Proof of Corollary 1

The proof is the same as that of Theorem 2 if we use

$$\psi(h) = \inf_{(z, x) \in \mathbb{H}_0} \psi_{z, x}(h); \psi_{z, x}(h) = \Pr[(Z, W) \in \{(\tilde{z}, \tilde{x}) \mid w_0 \|\tilde{z} - z\| + (1 - w_0) \|\tilde{x} - x\| \leq h\}],$$

to replace of the corresponding notation in Theorem 2.

□

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