

Supplementary Material to “Model Averaging for Estimating Treatment Effects”

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Appendix

A.1 Proof of Theorem 1

Let $\mathbf{b} = (b_1, \dots, b_{n_2})'$ and $\boldsymbol{\eta} = (\eta_1, \dots, \eta_{n_2})'$. Then

$$\begin{aligned} C_n(\mathbf{w}) &= \left\| \tilde{\mathbf{Y}} - \widehat{\boldsymbol{\Delta}}(\mathbf{w}) \right\|^2 + 2\sigma_t^2 \text{tr}(\mathbf{P}_{t_0}(\mathbf{w})) + 2\sigma_c^2 \text{tr}(\mathbf{P}_{c_0}(\mathbf{w})) \\ &= \left\| \boldsymbol{\Delta} + \mathbf{b} + \boldsymbol{\eta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w}) \right\|^2 + 2\sigma_t^2 \text{tr}(\mathbf{P}_{t_0}(\mathbf{w})) + 2\sigma_c^2 \text{tr}(\mathbf{P}_{c_0}(\mathbf{w})) \\ &= 2 \left[\langle \boldsymbol{\Delta} + \mathbf{b} - \widehat{\boldsymbol{\Delta}}(\mathbf{w}), \boldsymbol{\eta} \rangle + \sigma_t^2 \text{tr}(\mathbf{P}_{t_0}(\mathbf{w})) + \sigma_c^2 \text{tr}(\mathbf{P}_{c_0}(\mathbf{w})) \right] \\ &\quad + \left\| \boldsymbol{\Delta} + \mathbf{b} - \widehat{\boldsymbol{\Delta}}(\mathbf{w}) \right\|^2 + \|\boldsymbol{\eta}\|^2 \\ &:= \underline{A}(\mathbf{w}) + \underline{B}(\mathbf{w}) + \|\boldsymbol{\eta}\|^2. \end{aligned} \tag{A.1}$$

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Let us first consider the second term on the right-hand-side of (A.1)

$$\underline{B}(\mathbf{w}) = \left\| \boldsymbol{\Delta} + \mathbf{b} - \widehat{\boldsymbol{\Delta}}(\mathbf{w}) \right\|^2 = \|\mathbf{b}\|^2 + 2\langle \boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w}), \mathbf{b} \rangle + \left\| \boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w}) \right\|^2.$$

From (11) and Condition (14), we have

$$\begin{aligned} \frac{E(\|\mathbf{b}\|^2 | \mathbf{U})}{E(\|\boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w})\|^2 | \mathbf{U})} &\leq E(\|\mathbf{b}\|^2 | \mathbf{U}) / \xi_n \\ &= n_2 O\left(\left(p\left(\frac{\log n}{n}\right)^{1/p}\right)^2\right) / \xi_n \\ &= O\left(\frac{n}{\log n}\right) O\left(p^2\left(\frac{\log n}{n}\right)^{2/p}\right) / \xi_n \\ &= O\left(p^2\left(\frac{n}{\log n}\right)^{1-(2/p)} \xi_n^{-1}\right) \\ &= o(1), \quad a.s. \end{aligned} \tag{A.2}$$

By (A.2) and the Cauchy-Schwarz inequality, we obtain

$$\begin{aligned} &E(\langle \boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w}), \mathbf{b} \rangle | \mathbf{U}) \\ &\leq E\left[\left(\|\boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w})\|^2 \|\mathbf{b}\|^2\right)^{1/2} | \mathbf{U}\right] \\ &= E\left[\left(\|\boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w})\| \|\mathbf{b}\|\right) | \mathbf{U}\right] \\ &\leq \left\{E\left(\|\boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w})\|^2 | \mathbf{U}\right) E(\|\mathbf{b}\|^2 | \mathbf{U})\right\}^{1/2} \\ &= E\left(\|\boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w})\|^2 | \mathbf{U}\right) \left\{\frac{E(\|\mathbf{b}\|^2 | \mathbf{U})}{E\left(\|\boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w})\|^2 | \mathbf{U}\right)}\right\}^{1/2} \\ &= R_n(\mathbf{w}) o(1), \quad a.s. \end{aligned} \tag{A.3}$$

Then

$$\begin{aligned} E(\underline{B}(\mathbf{w}) | \mathbf{U}) &= E\left[\left\|\boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w})\right\|^2 + 2\langle \boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w}), \mathbf{b} \rangle + \|\mathbf{b}\|^2 | \mathbf{U}\right] \\ &\leq R_n(\mathbf{w}) + E\left(\|\boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w})\|^2 | \mathbf{U}\right) \left\{\frac{E(\|\mathbf{b}\|^2 | \mathbf{U})}{E\left(\|\boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w})\|^2 | \mathbf{U}\right)}\right\}^{1/2} \end{aligned}$$

$$\begin{aligned}
& +E\left(\left\|\Delta - \widehat{\Delta}(\mathbf{w})\right\|^2 \mid \mathbf{U}\right) \frac{E\left(\|\mathbf{b}\|^2 \mid \mathbf{U}\right)}{E\left(\left\|\Delta - \widehat{\Delta}(\mathbf{w})\right\|^2 \mid \mathbf{U}\right)} \\
& = R_n(\mathbf{w}) + R_n(\mathbf{w}) o(1) + R_n(\mathbf{w}) o(1) \\
& = R_n(\mathbf{w}) (1 + o(1)), \quad a.s.
\end{aligned} \tag{A.4}$$

Let $\zeta_{ta} = \mathbf{Y}_{ta} - \mathbf{f}_{ta}$, $\mathbf{v}_{ca} = \mathbf{Y}_{ca} - \mathbf{f}_{ca}$, $\widehat{\mathbf{f}}_t^{(k)} = \mathbf{P}_t^{(k)} \mathbf{Y}_{ta}$ and $\widehat{\mathbf{f}}_c^{(k)} = \mathbf{P}_{\bar{c}}^{(k)} \mathbf{Y}_{ca}$. From (12), we obtain the model average estimator

$$\begin{aligned}
\widehat{\Delta}(\mathbf{w}) & = \sum_{k=1}^K w_k \left(\widehat{\mathbf{f}}_t^{(k)} - \widehat{\mathbf{f}}_c^{(k)} \right) \\
& = \sum_{k=1}^K w_k \left[\mathbf{P}_t^{(k)} (\mathbf{f}_{ta} + \zeta_{ta}) - \mathbf{P}_{\bar{c}}^{(k)} (\mathbf{f}_{ca} + \mathbf{v}_{ca}) \right] \\
& = \mathbf{P}_t(\mathbf{w}) \mathbf{f}_{ta} + \mathbf{P}_t(\mathbf{w}) \zeta_{ta} - \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{f}_{ca} - \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca},
\end{aligned}$$

which yields

$$\begin{aligned}
\Delta + \mathbf{b} - \widehat{\Delta}(\mathbf{w}) & = \mathbf{f}_t - \mathbf{f}_c - \widehat{\Delta}(\mathbf{w}) \\
& = (\mathbf{f}_t - \mathbf{P}_t(\mathbf{w}) \mathbf{f}_{ta}) - (\mathbf{f}_c - \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{f}_{ca}) - \mathbf{P}_t(\mathbf{w}) \zeta_{ta} + \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca} \\
& := \mathbf{A}_t(\mathbf{w}) - \mathbf{A}_{\bar{c}}(\mathbf{w}) - \mathbf{P}_t(\mathbf{w}) \zeta_{ta} + \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca}.
\end{aligned} \tag{A.5}$$

Let $\zeta_t = (\zeta_1^t, \dots, \zeta_{n_2}^t)'$ and $\mathbf{v}_c = (v_1^t, \dots, v_{n_2}^t)'$. From (3) and (4), we have

$$E((\eta_m)^2 \mid \mathbf{U}) = \sigma_t^2 + \sigma_c^2 \quad \text{and} \quad E(\zeta_t' \mathbf{v}_c \mid \mathbf{U}) = 0. \tag{A.6}$$

Then from (A.5) and (A.6), we obtain

$$\begin{aligned}
E(\underline{A}(\mathbf{w}) \mid \mathbf{U}) & = 2E\left[\left\{\left\langle \mathbf{A}_t(\mathbf{w}) - \mathbf{A}_{\bar{c}}(\mathbf{w}) - \mathbf{P}_t(\mathbf{w}) \zeta_{ta} + \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca}, \zeta_t - \mathbf{v}_c \right\rangle \right. \right. \\
& \quad \left. \left. + \sigma_t^2 \text{tr}(\mathbf{P}_{t0}(\mathbf{w})) + \sigma_c^2 \text{tr}(\mathbf{P}_{\bar{c}0}(\mathbf{w})) \right\} \mid \mathbf{U}\right] \\
& = 2E\left\{\left\langle \mathbf{A}_t(\mathbf{w}) - \mathbf{A}_{\bar{c}}(\mathbf{w}), \boldsymbol{\eta} \right\rangle \mid \mathbf{U}\right\} + 2E\left\{\left\langle -\mathbf{P}_t(\mathbf{w}) \zeta_{ta}, \zeta_t \right\rangle \mid \mathbf{U}\right\} \\
& \quad + 2\sigma_t^2 \text{tr}(\mathbf{P}_{t0}(\mathbf{w})) + 2E\left(\left\langle \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca}, -\mathbf{v}_c \right\rangle \mid \mathbf{U}\right) + 2\sigma_t^2 \text{tr}(\mathbf{P}_{\bar{c}}(\mathbf{w})) \\
& \quad + 2E\left(\left\langle -\mathbf{P}_t(\mathbf{w}) \zeta_{ta}, -\mathbf{v}_c \right\rangle \mid \mathbf{U}\right) + 2E\left(\left\langle \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca}, \zeta_t \right\rangle \mid \mathbf{U}\right) \\
& = 0
\end{aligned} \tag{A.7}$$

and

$$E(\|\boldsymbol{\eta}\|^2 \mid \mathbf{U}) = n_2(\sigma_t^2 + \sigma_c^2). \tag{A.8}$$

Now, from (A.1), (A.4), (A.7) and (A.8), we have

$$\begin{aligned}
E(C_n(\mathbf{w}) \mid \mathbf{U}) & = E(\underline{A}(\mathbf{w}) \mid \mathbf{U}) + E(\underline{B}(\mathbf{w}) \mid \mathbf{U}) + E(\|\boldsymbol{\eta}\|^2 \mid \mathbf{U}) \\
& = R_n(\mathbf{w}) (1 + o(1)) + n_2(\sigma_t^2 + \sigma_c^2), \quad a.s.
\end{aligned} \tag{A.9}$$

This completes the proof of Theorem 1. \square

A.2 Proof of Theorem 2

The proof of Theorem 2 is an application of Whittle (1960) and Chebyshev's inequality. Following Zhang et al. (2013) proof of their Theorem 2.1, we assume, for purposes of convenience, that \mathbf{U} is non-stochastic. The proof that follows also applies to the case of stochastic \mathbf{U} because all of the technical conditions imposed on \mathbf{U} hold almost surely. In the following proof, we assume that $C_1, \dots, C_{15}, C'_8, \dots$, and C'_{13} are distinct constants.

Now, from (A.1), it follows that

$$\begin{aligned} C_n(\mathbf{w}) &= L_n(\mathbf{w}) + \underline{A}(\mathbf{w}) + (\underline{B}(\mathbf{w}) - L_n(\mathbf{w})) + \|\boldsymbol{\eta}\|^2 \\ &=: L_n(\mathbf{w}) + \underline{A}(\mathbf{w}) + \underline{B}(\mathbf{w}) + \|\boldsymbol{\eta}\|^2. \end{aligned} \quad (\text{A.10})$$

Theorem 2 is valid if the following hold

$$\sup_{\mathbf{w} \in H_n} |\underline{A}(\mathbf{w})|/R_n(\mathbf{w}) \rightarrow_p 0, \quad (\text{A.11})$$

$$\sup_{\mathbf{w} \in H_n} |\underline{B}(\mathbf{w})/R_n(\mathbf{w})| \rightarrow_p 0 \quad (\text{A.12})$$

and

$$\sup_{\mathbf{w} \in H_n} |L_n(\mathbf{w})/R_n(\mathbf{w}) - 1| \rightarrow_p 0. \quad (\text{A.13})$$

From (A.5), we have

$$\begin{aligned} \underline{A}(\mathbf{w}) &= 2\langle \boldsymbol{\Delta} + \mathbf{b} - \widehat{\boldsymbol{\Delta}}(\mathbf{w}), \boldsymbol{\eta} \rangle + 2\sigma_t^2 \text{tr}(\mathbf{P}_t(\mathbf{w})) + 2\sigma_c^2 \text{tr}(\mathbf{P}_{\bar{c}}(\mathbf{w})) \\ &= 2\langle \mathbf{f}_t - \mathbf{f}_c - \mathbf{P}_t(\mathbf{w})\mathbf{f}_{ta} + \mathbf{P}_{\bar{c}}(\mathbf{w})\mathbf{f}_{ca} - \mathbf{P}_t(\mathbf{w})\boldsymbol{\zeta}_{ta} + \mathbf{P}_{\bar{c}}(\mathbf{w})\mathbf{v}_{ca}, \boldsymbol{\eta} \rangle \\ &\quad + 2\sigma_t^2 \text{tr}(\mathbf{P}_{t0}(\mathbf{w})) + 2\sigma_c^2 \text{tr}(\mathbf{P}_{\bar{c}0}(\mathbf{w})) \\ &= 2\langle \mathbf{f}_t - \mathbf{P}_t(\mathbf{w})\mathbf{f}_{ta}, \boldsymbol{\eta} \rangle - 2\langle \mathbf{f}_c - \mathbf{P}_{\bar{c}}(\mathbf{w})\mathbf{f}_{ca}, \boldsymbol{\eta} \rangle - 2\{\langle \mathbf{P}_t(\mathbf{w})\boldsymbol{\zeta}_{ta}, \boldsymbol{\eta} \rangle \\ &\quad - \sigma_t^2 \text{tr}(\mathbf{P}_{t0}(\mathbf{w}))\} + 2\{\langle \mathbf{P}_{\bar{c}0}(\mathbf{w})\mathbf{v}_{ca}, \boldsymbol{\eta} \rangle + \sigma_c^2 \text{tr}(\mathbf{P}_{\bar{c}0}(\mathbf{w}))\} \\ &= 2\langle \mathbf{A}_t(\mathbf{w}), \boldsymbol{\eta} \rangle - 2\langle \mathbf{A}_{\bar{c}}(\mathbf{w}), \boldsymbol{\eta} \rangle - 2\{\langle \mathbf{P}_t(\mathbf{w})\boldsymbol{\zeta}_{ta}, \boldsymbol{\eta} \rangle - \sigma_t^2 \text{tr}(\mathbf{P}_{t0}(\mathbf{w}))\} \\ &\quad + 2\{\langle \mathbf{P}_{\bar{c}}(\mathbf{w})\mathbf{v}_{ca}, \boldsymbol{\eta} \rangle + \sigma_c^2 \text{tr}(\mathbf{P}_{\bar{c}0}(\mathbf{w}))\}. \end{aligned} \quad (\text{A.14})$$

Hence (A.3) is valid if the following conditions hold

$$\sup_{\mathbf{w} \in H_n} |\langle \mathbf{A}_t(\mathbf{w}), \boldsymbol{\eta} \rangle|/R_n(\mathbf{w}) \rightarrow_p 0, \quad (\text{A.15})$$

$$\sup_{\mathbf{w} \in H_n} |\langle \mathbf{A}_{\bar{c}}(\mathbf{w}), \boldsymbol{\eta} \rangle|/R_n(\mathbf{w}) \rightarrow_p 0, \quad (\text{A.16})$$

$$\sup_{\mathbf{w} \in H_n} |\langle \mathbf{P}_t(\mathbf{w})\boldsymbol{\zeta}_{ta}, \boldsymbol{\eta} \rangle - \sigma_t^2 \text{tr}(\mathbf{P}_{t0}(\mathbf{w}))|/R_n(\mathbf{w}) \rightarrow_p 0 \quad (\text{A.17})$$

and

$$\sup_{\mathbf{w} \in H_n} |\langle \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca}, \boldsymbol{\eta} \rangle + \sigma_c^2 \text{tr}(\mathbf{P}_{\bar{c}0}(\mathbf{w}))| / R_n(\mathbf{w}) \rightarrow_p 0. \quad (\text{A.18})$$

Note that Condition (17) implies that

$$E((\zeta_m^t)^{4G} | \mathbf{U}) \leq \kappa < \infty \text{ and } E((v_m^c)^{4G} | \mathbf{U}) \leq \kappa < \infty, m = 1, \dots, n_2. \quad (\text{A.19})$$

To prove (A.15), using (18), (A.19), Chebyshev’s inequality and Theorem 2 of Whittle (1960), and following steps similar to the proof of Equation (A.1) in Wan et al. (2010), we have, for any $\delta > 0$,

$$\begin{aligned} & P \left\{ \sup_{\mathbf{w} \in H_n} |\langle \mathbf{A}_t(\mathbf{w}), \boldsymbol{\eta} \rangle| / R_n(\mathbf{w}) > \delta \right\} \\ & \leq P \left\{ \sup_{\mathbf{w} \in H_n} \sum_{k=1}^K w_k \left| \boldsymbol{\eta}' (\mathbf{f}_t - \mathbf{P}_t^{(k)} \mathbf{f}_{ta}) \right| > \delta \xi_n \right\} \\ & \leq P \left\{ \max_{1 \leq k \leq K} \left| \boldsymbol{\eta}' (\mathbf{f}_t - \mathbf{P}_t^{(k)} \mathbf{f}_{ta}) \right| > \delta \xi_n \right\} \\ & = P \left\{ \{ |\langle \boldsymbol{\eta}, \mathbf{A}_t(\mathbf{w}_1^0) \rangle| > \delta \xi_n \} \cup \{ |\langle \boldsymbol{\eta}, \mathbf{A}_t(\mathbf{w}_2^0) \rangle| > \delta \xi_n \} \right. \\ & \quad \left. \cup \dots \cup \{ |\langle \boldsymbol{\eta}, \mathbf{A}_t(\mathbf{w}_K^0) \rangle| > \delta \xi_n \} \right\} \\ & \leq \sum_{k=1}^K P \{ |\langle \boldsymbol{\eta}, \mathbf{A}_t(\mathbf{w}_k^0) \rangle| > \delta \xi_n \} \\ & \leq \sum_{k=1}^K E \left\{ \frac{|\langle \boldsymbol{\eta}, \mathbf{A}_t(\mathbf{w}_k^0) \rangle|^{2G}}{\delta^{2G} \xi_n^{2G}} \right\} \\ & \leq C_1 \delta^{-2G} \xi_n^{-2G} \sum_{k=1}^K \|\mathbf{A}_t(\mathbf{w}_k^0)\|^{2G} \rightarrow 0. \end{aligned}$$

Similarly, for (A.16), we obtain

$$P \left\{ \sup_{\mathbf{w} \in H_n} |\langle \mathbf{A}_{\bar{c}}(\mathbf{w}), \boldsymbol{\eta} \rangle| / R_n(\mathbf{w}) > \delta \right\} \leq C_2 \delta^{-2G} \xi_n^{-2G} \sum_{k=1}^K \|\mathbf{A}_{\bar{c}}(\mathbf{w}_k^0)\|^{2G} \rightarrow 0.$$

Now, let us prove (A.17). As

$$\begin{aligned} & P \left\{ \sup_{\mathbf{w} \in H_n} |\langle \mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta}, \boldsymbol{\eta} \rangle - \sigma_t^2 \text{tr}(\mathbf{P}_{t0}(\mathbf{w}))| / R_n(\mathbf{w}) > \delta \right\} \\ & = P \left\{ \sup_{\mathbf{w} \in H_n} |\langle \mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta}, \boldsymbol{\zeta}_t - \mathbf{v}_c \rangle - \sigma_t^2 \text{tr}(\mathbf{P}_{t0}(\mathbf{w}))| / R_n(\mathbf{w}) > \delta \right\} \\ & \leq P \left\{ \sup_{\mathbf{w} \in H_n} |\langle \mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta}, \boldsymbol{\zeta}_t \rangle - \sigma_t^2 \text{tr}(\mathbf{P}_{t0}(\mathbf{w}))| / R_n(\mathbf{w}) > \delta \right\} \\ & \quad + P \left\{ \sup_{\mathbf{w} \in H_n} |\langle \mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta}, \mathbf{v}_c \rangle| / R_n(\mathbf{w}) > \delta \right\}, \quad (\text{A.20}) \end{aligned}$$

it suffices to show that the two terms on the right hand side of (A.20) approach zero. Note that the first term is bounded by

$$\begin{aligned}
& P \left\{ \sup_{\mathbf{w} \in H_n} |\langle \mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta}, \boldsymbol{\zeta}_t \rangle - \sigma_t^2 \text{tr}(\mathbf{P}_{t0}(\mathbf{w}))| / R_n(\mathbf{w}) > \delta \right\} \\
& \leq \sum_{k=1}^K P \left\{ |\langle \mathbf{P}_t(\mathbf{w}_k^0) \boldsymbol{\zeta}_{ta}, \boldsymbol{\zeta}_t \rangle - \sigma_t^2 \text{tr}(\mathbf{P}_{t0}(\mathbf{w}_k^0))| > \delta \xi_n \right\} \\
& \leq \sum_{k=1}^K P \left\{ |\langle \mathbf{P}_{t0}(\mathbf{w}_k^0) \boldsymbol{\zeta}_{ta}, \boldsymbol{\zeta}_{ta} \rangle - \sigma_t^2 \text{tr}(\mathbf{P}_{t0}(\mathbf{w}_k^0))| > \delta \xi_n \right\} \\
& \leq \sum_{k=1}^K E \left[\frac{\{ \langle \mathbf{P}_{t0}(\mathbf{w}_k^0) \boldsymbol{\zeta}_{ta}, \boldsymbol{\zeta}_{ta} \rangle - \sigma_t^2 \text{tr}(\mathbf{P}_{t0}(\mathbf{w}_k^0)) \}^2}{\delta^2 \xi_n^{2G}} \right] \\
& \leq C_3 \delta^{-2G} \xi_n^{-2G} \sum_{k=1}^K [\text{tr}(\mathbf{P}_{t0}(\mathbf{w}_k^0) \mathbf{P}'_{t0}(\mathbf{w}_k^0))]^G \\
& = C_3 \delta^{-2G} \xi_n^{-2G} \sum_{k=1}^K [\text{tr}(\mathbf{P}_t(\mathbf{w}_k^0) \mathbf{P}'_t(\mathbf{w}_k^0))]^G \\
& \rightarrow 0,
\end{aligned}$$

while the second term is given by

$$\begin{aligned}
& P \left\{ \sup_{\mathbf{w} \in H_n} |\langle \mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta}, \mathbf{v}_c \rangle| / R_n(\mathbf{w}) > \delta \right\} \\
& \leq \sum_{k=1}^K P \left\{ |\langle \mathbf{P}_t(\mathbf{w}_k^0) \boldsymbol{\zeta}_{ta}, \mathbf{v}_c \rangle - E \langle \mathbf{P}_t(\mathbf{w}_k^0) \boldsymbol{\zeta}_{ta}, \mathbf{v}_c \rangle| > \delta \xi_n \right\} \\
& = \sum_{k=1}^K P \left[|\langle \mathbf{P}_t(\mathbf{w}_k^0) \boldsymbol{\zeta}_{ta}, \mathbf{v}_{ca} \rangle| > \delta \xi_n \right] \\
& \leq \sum_{k=1}^K E \left[\frac{\{ \langle \mathbf{P}_t(\mathbf{w}_k^0) \boldsymbol{\zeta}_{ta}, \mathbf{v}_c \rangle \}^2}{\delta^2 \xi_n^2} \right] \\
& \leq \sum_{k=1}^K E \left[\frac{\{ \langle \boldsymbol{\zeta}'_{ta} \mathbf{P}'_t(\mathbf{w}_k^0) \mathbf{v}_c \mathbf{v}'_c \mathbf{P}_t(\mathbf{w}_k^0) \boldsymbol{\zeta}_{ta} \rangle \}}{\delta^2 \xi_n^2} \right] \\
& \leq C_4 \delta^{-2} \xi_n^{-2} \sum_{k=1}^K [\text{tr}(\mathbf{P}_t(\mathbf{w}_k^0) \mathbf{P}'_t(\mathbf{w}_k^0))] \\
& \rightarrow 0. \tag{A.21}
\end{aligned}$$

Hence (A.17) is valid.

Similarly, for (A.18), it is observed that

$$P \left\{ \sup_{\mathbf{w} \in H_n} |\langle \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca}, \boldsymbol{\eta} \rangle + \sigma_c^2 \text{tr}(\mathbf{P}_{\bar{c}}(\mathbf{w}))| / R_n(\mathbf{w}) > \delta \right\}$$

$$\begin{aligned}
&= P \left\{ \sup_{\mathbf{w} \in H_n} \left| \langle \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca}, \boldsymbol{\zeta}_t - \mathbf{v}_c \rangle + \sigma_c^2 \text{tr}(\mathbf{P}_{\bar{c}}(\mathbf{w})) \right| / R_n(\mathbf{w}) > \delta \right\} \\
&\leq P \left\{ \sup_{\mathbf{w} \in H_n} \left| \langle \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca}, \mathbf{v}_c \rangle - \sigma_c^2 \text{tr}(\mathbf{P}_{\bar{c}}(\mathbf{w})) \right| / R_n(\mathbf{w}) > \delta \right\} \\
&\quad + P \left\{ \sup_{\mathbf{w} \in H_n} \left| \langle \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca}, \boldsymbol{\zeta}_t \rangle \right| / R_n(\mathbf{w}) > \delta \right\} \\
&\leq C_5 \delta^{-2G} \xi_n^{-2G} \sum_{k=1}^K [\text{tr}(\mathbf{P}'_{\bar{c}}(\mathbf{w}_k^0) \mathbf{P}_{\bar{c}}(\mathbf{w}_k^0))]^G + C_6 \delta^{-2} \xi_n^{-2} \sum_{k=1}^K [\text{tr}(\mathbf{P}'_{\bar{c}}(\mathbf{w}_k^0) \mathbf{P}_{\bar{c}}(\mathbf{w}_k^0))] \\
&\rightarrow 0. \tag{A.22}
\end{aligned}$$

Given (A.15)-(A.18), it is straightforward to see that (A.3) is valid.

Now, to prove (A.12), let us first consider

$$\begin{aligned}
\underline{B}(\mathbf{w}) &= \left\| \boldsymbol{\Delta} + \mathbf{b} - \widehat{\boldsymbol{\Delta}}(\mathbf{w}) \right\|^2 \\
&= \left\| \mathbf{A}_t(\mathbf{w}) - \mathbf{A}_{\bar{c}}(\mathbf{w}) \right\|^2 + \left\| \mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta} - \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca} \right\|^2 \\
&\quad - 2 \langle \mathbf{A}_t(\mathbf{w}) - \mathbf{A}_{\bar{c}}(\mathbf{w}), \mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta} - \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca} \rangle. \tag{A.23}
\end{aligned}$$

Define

$$\begin{aligned}
\tilde{R}_n(\mathbf{w}) &:= E(\underline{B}(\mathbf{w}) | \mathbf{U}) = E \left[\left\| \mathbf{A}_t(\mathbf{w}) - \mathbf{A}_{\bar{c}}(\mathbf{w}) \right\|^2 + \left\| \mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta} - \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca} \right\|^2 \right. \\
&\quad \left. - 2 \langle \mathbf{A}_t(\mathbf{w}) - \mathbf{A}_{\bar{c}}(\mathbf{w}), \mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta} - \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca} \rangle \middle| \mathbf{U} \right] \\
&= \left\| \mathbf{A}_t(\mathbf{w}) - \mathbf{A}_{\bar{c}}(\mathbf{w}) \right\|^2 + \sigma_t^2 \text{tr}(\mathbf{P}'_{\bar{c}}(\mathbf{w}) \mathbf{P}_{\bar{c}}(\mathbf{w})) \\
&\quad + \sigma_t^2 \text{tr}(\mathbf{P}'_t(\mathbf{w}) \mathbf{P}_t(\mathbf{w})). \tag{A.24}
\end{aligned}$$

Then from (A.2) and (A.3), we obtain

$$\begin{aligned}
&\sup_{\mathbf{w} \in H_n} \left| \tilde{R}_n(\mathbf{w}) - R_n(\mathbf{w}) \right| / R_n(\mathbf{w}) \\
&= \sup_{\mathbf{w} \in H_n} E \left[\left(\left\| \boldsymbol{\Delta} + \mathbf{b} - \widehat{\boldsymbol{\Delta}}(\mathbf{w}) \right\|^2 - \left\| \boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w}) \right\|^2 \right) \middle| \mathbf{U} \right] / R_n(\mathbf{w}) \\
&= \sup_{\mathbf{w} \in H_n} E \left[2 \left(\langle \boldsymbol{\Delta} - \widehat{\boldsymbol{\Delta}}(\mathbf{w}), \mathbf{b} \rangle + \|\mathbf{b}\|^2 \right) \middle| \mathbf{U} \right] / R_n(\mathbf{w}) \\
&= o(1). \tag{A.25}
\end{aligned}$$

By (A.23)-(A.25), we have

$$\begin{aligned}
&\sup_{\mathbf{w} \in H_n} |\underline{B}(\mathbf{w}) / R_n(\mathbf{w}) - 1| \\
&= \sup_{\mathbf{w} \in H_n} \left[\left\{ \underline{B}(\mathbf{w}) - \tilde{R}_n(\mathbf{w}) \right\} + \left\{ \tilde{R}_n(\mathbf{w}) - R_n(\mathbf{w}) \right\} \right] / R_n(\mathbf{w}) \\
&= \sup_{\mathbf{w} \in H_n} \left\{ \left\| \mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta} - \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca} \right\|^2 - \sigma_c^2 \text{tr}(\mathbf{P}'_{\bar{c}}(\mathbf{w}) \mathbf{P}_{\bar{c}}(\mathbf{w})) - \sigma_t^2 \text{tr}(\mathbf{P}'_t(\mathbf{w}) \mathbf{P}_t(\mathbf{w})) \right\}
\end{aligned}$$

$$\begin{aligned}
& -2\langle \mathbf{A}_t(\mathbf{w}) - \mathbf{A}_{\bar{c}}(\mathbf{w}), \mathbf{P}_t(\mathbf{w})\boldsymbol{\zeta}_{ta} - \mathbf{P}_{\bar{c}}(\mathbf{w})\mathbf{v}_{ca} \rangle \Big\} / R_n(\mathbf{w}) + o(1) \\
= & \sup_{\mathbf{w} \in H_n} \left[\left\{ \|\mathbf{P}_t(\mathbf{w})\boldsymbol{\zeta}_{ta}\|^2 - \sigma_t^2 \text{tr}(\mathbf{P}'_t(\mathbf{w})\mathbf{P}_t(\mathbf{w})) \right\} - 2\langle \mathbf{P}_t(\mathbf{w})\boldsymbol{\zeta}_{ta}, \mathbf{P}_{\bar{c}}(\mathbf{w})\mathbf{v}_{ca} \rangle \right. \\
& + \left\{ \|\mathbf{P}_{\bar{c}}(\mathbf{w})\mathbf{v}_{ca}\|^2 - \sigma_c^2 \text{tr}(\mathbf{P}'_{\bar{c}}(\mathbf{w})\mathbf{P}_{\bar{c}}(\mathbf{w})) \right\} - 2\langle \mathbf{A}_t(\mathbf{w}), \mathbf{P}_t(\mathbf{w})\boldsymbol{\zeta}_{ta} \rangle \\
& + 2\langle \mathbf{A}_t(\mathbf{w}), \mathbf{P}_{\bar{c}}(\mathbf{w})\mathbf{v}_{ca} \rangle + 2\langle \mathbf{A}_{\bar{c}}(\mathbf{w}), \mathbf{P}_t(\mathbf{w})\boldsymbol{\zeta}_{ta} \rangle \\
& \left. - 2\langle \mathbf{A}_{\bar{c}}(\mathbf{w}), \mathbf{P}_{\bar{c}}(\mathbf{w})\mathbf{v}_{ca} \rangle \right] / R_n(\mathbf{w}) + o(1). \tag{A.26}
\end{aligned}$$

Next, we will show that the terms on the right hand side of (A.26) converge to zero in probability. Denote the maximum singular value by $\bar{\lambda}$. Following Wan et al. (2010), we have

$$\begin{aligned}
\bar{\lambda}(\mathbf{P}_t^{(k)}) &= \bar{\lambda}(\mathbf{U}_t^{(k)}(\mathbf{U}_{ta}^{(k)'}\mathbf{U}_{ta}^{(k)})^{-1}\mathbf{U}_{ta}^{(k)'}) \\
&= \bar{\lambda}^{1/2}(\mathbf{U}_t^{(k)}(\mathbf{U}_{ta}^{(k)'}\mathbf{U}_{ta}^{(k)})^{-1}\mathbf{U}_{ta}^{(k)'}\mathbf{U}_{ta}^{(k)}(\mathbf{U}_{ta}^{(k)'}\mathbf{U}_{ta}^{(k)})^{-1}\mathbf{U}_t^{(k)'}) \\
&= \bar{\lambda}(\mathbf{U}_t^{(k)}(\mathbf{U}_{ta}^{(k)'}\mathbf{U}_{ta}^{(k)})^{-1}\mathbf{U}_t^{(k)'}) \\
&= \bar{\lambda}(\mathbf{U}_t^{(k)}(\mathbf{U}_t^{(k)'}\mathbf{U}_t^{(k)} + \mathbf{F})^{-1}\mathbf{U}_t^{(k)'}) \quad (\exists \mathbf{F} \geq 0) \\
&\leq 1.
\end{aligned}$$

Recognising that $\bar{\lambda}(\mathbf{P}_t(\mathbf{w}^*)) \leq \max_{1 \leq k \leq K} \bar{\lambda}(\mathbf{P}_t^{(k)}) \leq 1$ for any $\mathbf{w}, \mathbf{w}^* \in H_n$, we have

$$\begin{aligned}
\text{tr}\{\mathbf{P}'_t(\mathbf{w})\mathbf{P}_t(\mathbf{w}^*)\mathbf{P}'_t(\mathbf{w}^*)\mathbf{P}_t(\mathbf{w})\} &\leq \bar{\lambda}^2(\mathbf{P}_t(\mathbf{w}^*))\text{tr}\{\mathbf{P}_t(\mathbf{w})\mathbf{P}'_t(\mathbf{w})\} \\
&\leq \text{tr}\{\mathbf{P}_t(\mathbf{w})\mathbf{P}'_t(\mathbf{w})\}.
\end{aligned}$$

Furthermore, note that

$$\begin{aligned}
\bar{\lambda}(\mathbf{P}_{\bar{c}}(\mathbf{w}^*)) &\leq \bar{\lambda}(\mathbf{P}_c(\mathbf{w}^*)) + \bar{\lambda}(\mathbf{P}_{\bar{c}}(\mathbf{w}^*) - \mathbf{P}_c(\mathbf{w}^*)) \\
&\leq 1 + \max_{1 \leq k \leq K} \bar{\lambda}^{1/2} \left\{ \left(\mathbf{U}_t^{(k)} - \mathbf{U}_c^{(k)} \right) (\mathbf{U}_{ca}^{(k)'} \mathbf{U}_{ca}^{(k)})^{-1} \mathbf{U}_{ca}^{(k)'} \right\} \\
&\quad \left\{ \left(\mathbf{U}_t^{(k)} - \mathbf{U}_c^{(k)} \right) (\mathbf{U}_{ca}^{(k)'} \mathbf{U}_{ca}^{(k)})^{-1} \mathbf{U}_{ca}^{(k)'} \right\}' \\
&= 1 + \max_{k=1, \dots, K} \bar{\lambda}^{1/2} \left\{ \left(\mathbf{U}_t^{(k)} - \mathbf{U}_c^{(k)} \right) (\mathbf{U}_{ca}^{(k)'} \mathbf{U}_{ca}^{(k)})^{-1} \left(\mathbf{U}_t^{(k)} - \mathbf{U}_c^{(k)} \right)' \right\} \\
&\leq 1 + \max_{1 \leq k \leq K} \bar{\lambda}^{1/2} \left\{ \left(n_c^{-1} \mathbf{U}_{ca}^{(k)'} \mathbf{U}_{ca}^{(k)} \right)^{-1} \right\} \\
&\quad \times \max_{1 \leq k \leq K} \bar{\lambda}^{1/2} \left\{ n_c^{-1} \left(\mathbf{U}_t^{(k)} - \mathbf{U}_c^{(k)} \right)' \left(\mathbf{U}_t^{(k)} - \mathbf{U}_c^{(k)} \right) \right\} \\
&\leq 1 + C_0 \max_{1 \leq k \leq K} \bar{\lambda}^{1/2} \left\{ n_c^{-1} \left(\mathbf{U}_t^{(k)} - \mathbf{U}_c^{(k)} \right)' \left(\mathbf{U}_t^{(k)} - \mathbf{U}_c^{(k)} \right) \right\} \\
&\leq 1 + C_0 \max_{1 \leq k \leq K} \text{tr}^{1/2} \left\{ n_c^{-1} \left(\mathbf{U}_t^{(k)} - \mathbf{U}_c^{(k)} \right)' \left(\mathbf{U}_t^{(k)} - \mathbf{U}_c^{(k)} \right) \right\}
\end{aligned}$$

$$\begin{aligned}
&= 1 + C_0 \max_{1 \leq k \leq K} \left[n_c^{-1} n_2 l_k O \left(\left(\frac{\log n}{n} \right)^{2/p} \right) \right]^{1/2} \\
&= 1 + C_0 \max_{1 \leq k \leq K} \left[\frac{p}{\log n} O \left(\left(\frac{\log n}{n} \right)^{2/p} \right) \right]^{1/2} \\
&= 1 + o(1).
\end{aligned}$$

Then

$$\begin{aligned}
&P \left\{ \sup_{\mathbf{w} \in H_n} \left| \|\mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta}\|^2 - \sigma_t^2 \text{tr}(\mathbf{P}_t'(\mathbf{w}) \mathbf{P}_t(\mathbf{w})) \right| / R_n(\mathbf{w}) > \delta \right\} \\
&\leq P \left\{ \sup_{\mathbf{w} \in H_n} \sum_{j=1}^K \sum_{k=1}^K w_j w_k \left| \boldsymbol{\zeta}'_{ta} \mathbf{P}_t^{(j)'} \mathbf{P}_t^{(k)} \boldsymbol{\zeta}_{ta} - \sigma_t^2 \text{tr}(\mathbf{P}_t^{(j)'} \mathbf{P}_t^{(k)}) \right| > \delta \xi_n \right\} \\
&\leq P \left\{ \max_{1 \leq j \leq K} \max_{1 \leq k \leq K} \left| \boldsymbol{\zeta}'_{ta} \mathbf{P}_t^{(j)'} \mathbf{P}_t^{(k)} \boldsymbol{\zeta}_{ta} - \sigma_t^2 \text{tr}(\mathbf{P}_t^{(j)'} \mathbf{P}_t^{(k)}) \right| > \delta \xi_n \right\} \\
&\leq \sum_{j=1}^K \sum_{k=1}^K P \left\{ \left| \langle \boldsymbol{\zeta}_{ta}, \mathbf{P}_t'(\mathbf{w}_j^0) \mathbf{P}_t(\mathbf{w}_k^0) \boldsymbol{\zeta}_{ta} \rangle - \sigma_t^2 \text{tr}(\mathbf{P}_t'(\mathbf{w}_j^0) \mathbf{P}_t(\mathbf{w}_k^0)) \right| > \delta \xi_n \right\} \\
&\leq \sum_{j=1}^K \sum_{k=1}^K E \left[\frac{\left\{ \langle \boldsymbol{\zeta}_{ta}, \mathbf{P}_t'(\mathbf{w}_j^0) \mathbf{P}_t(\mathbf{w}_k^0) \boldsymbol{\zeta}_{ta} \rangle - \sigma_t^2 \text{tr}(\mathbf{P}_t'(\mathbf{w}_j^0) \mathbf{P}_t(\mathbf{w}_k^0)) \right\}^{2G}}{\delta^{2G} \xi_n^{2G}} \right] \\
&\leq C_7 \delta^{-2G} \xi_n^{-2G} \sum_{j=1}^K \sum_{k=1}^K \left[\text{tr}(\mathbf{P}_t'(\mathbf{w}_k^0) \mathbf{P}_t(\mathbf{w}_j^0) \mathbf{P}_t'(\mathbf{w}_j^0) \mathbf{P}_t(\mathbf{w}_k^0)) \right]^G \\
&\leq C_7 \delta^{-2G} \xi_n^{-2G} \bar{\lambda}^2(\mathbf{P}_t(\mathbf{w}_j^0)) \sum_{j=1}^K \sum_{k=1}^K \left[\text{tr}(\mathbf{P}_t'(\mathbf{w}_k^0) \mathbf{P}_t(\mathbf{w}_k^0)) \right]^G \\
&\leq C_7 \delta^{-2G} \xi_n^{-2G} K \sum_{k=1}^K (R_{n_t}(\mathbf{w}_k^0))^G \\
&\rightarrow 0
\end{aligned}$$

and

$$\begin{aligned}
&P \left\{ \sup_{\mathbf{w} \in H_n} \left| \|\mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca}\|^2 - \sigma_{\bar{c}}^2 \text{tr}(\mathbf{P}_{\bar{c}}'(\mathbf{w}) \mathbf{P}_{\bar{c}}(\mathbf{w})) \right| / R_n(\mathbf{w}) > \delta \right\} \\
&\leq P \left\{ \sup_{\mathbf{w} \in H_n} \sum_{j=1}^K \sum_{k=1}^K w_j w_k \left| \mathbf{v}'_{ca} \mathbf{P}_{\bar{c}}^{(j)'} \mathbf{P}_{\bar{c}}^{(k)} \mathbf{v}_{ca} - \sigma_{\bar{c}}^2 \text{tr}(\mathbf{P}_{\bar{c}}^{(j)'} \mathbf{P}_{\bar{c}}^{(k)}) \right| > \delta \xi_n \right\} \\
&\leq P \left\{ \max_{1 \leq j \leq K} \max_{1 \leq k \leq K} \left| \mathbf{v}'_{ca} \mathbf{P}_{\bar{c}}^{(j)'} \mathbf{P}_{\bar{c}}^{(k)} \mathbf{v}_{ca} - \sigma_{\bar{c}}^2 \text{tr}(\mathbf{P}_{\bar{c}}^{(j)'} \mathbf{P}_{\bar{c}}^{(k)}) \right| > \delta \xi_n \right\} \\
&\leq C_8 \delta^{-2G} \xi_n^{-2G} \sum_{j=1}^K \sum_{k=1}^K \left[\text{tr} \left\{ \mathbf{P}_{\bar{c}}'(\mathbf{w}_k^0) \mathbf{P}_{\bar{c}}(\mathbf{w}_j^0) \mathbf{P}_{\bar{c}}'(\mathbf{w}_j^0) \mathbf{P}_{\bar{c}}(\mathbf{w}_k^0) \right\} \right]^G
\end{aligned}$$

$$\begin{aligned}
&\leq C_8 \delta^{-2G} \xi_n^{-2G} \bar{\lambda}(\mathbf{P}'_{\bar{c}}(\mathbf{w}^*) \mathbf{P}_{\bar{c}}(\mathbf{w}^*)) \sum_{j=1}^K \sum_{k=1}^K [\text{tr} \{ \mathbf{P}'_{\bar{c}}(\mathbf{w}_k^0) \mathbf{P}_{\bar{c}}(\mathbf{w}_k^0) \}]^G \\
&\leq C'_8 \delta^{-2G} \xi_n^{-2G} K \sum_{k=1}^K (R_{n_{\bar{c}}}(\mathbf{w}_k^0))^G \\
&\rightarrow 0.
\end{aligned}$$

Additionally, we can show that

$$\begin{aligned}
&P \left\{ \sup_{\mathbf{w} \in H_n} |\langle \mathbf{A}_t(\mathbf{w}), \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca} \rangle| / R_n(\mathbf{w}) > \delta \right\} \\
&\leq P \left\{ \sup_{\mathbf{w} \in H_n} \left| \mathbf{v}'_{ca} \sum_{j=1}^K w_j \mathbf{P}_{\bar{c}}^{(j)'} \sum_{k=1}^K w_k (\mathbf{f}_t - \mathbf{P}_t^{(k)} \mathbf{f}_{ta}) \right| > \delta \xi_n \right\} \\
&\leq P \left\{ \sum_{j=1}^K \sum_{k=1}^K w_j w_k \left| \mathbf{v}'_{ca} \mathbf{P}_{\bar{c}}^{(j)'} (\mathbf{f}_t - \mathbf{P}_t^{(k)} \mathbf{f}_{ta}) \right| > \delta \xi_n \right\} \\
&\leq P \left\{ \max_{1 \leq j \leq K} \max_{1 \leq k \leq K} \left| \mathbf{v}'_{ca} \mathbf{P}_{\bar{c}}^{(j)'} (\mathbf{f}_t - \mathbf{P}_t^{(k)} \mathbf{f}_{ta}) \right| > \delta \xi_n \right\} \\
&\leq \sum_{j=1}^K \sum_{k=1}^K P \left\{ |\langle \mathbf{A}_t(\mathbf{w}_j^0), \mathbf{P}_{\bar{c}}(\mathbf{w}_k^0) \mathbf{v}_{ca} \rangle| > \delta \xi_n \right\} \\
&\leq \sum_{j=1}^K \sum_{k=1}^K E \left[\frac{|\langle \mathbf{A}_t(\mathbf{w}_j^0), \mathbf{P}_{\bar{c}}(\mathbf{w}_k^0) \mathbf{v}_{ca} \rangle|^{2G}}{\delta^{2G} \xi_n^{2G}} \right] \\
&\leq C_9 \delta^{-2G} \xi_n^{-2G} \sum_{j=1}^K \sum_{k=1}^K \left\| \mathbf{P}_{\bar{c}}^{(j)'} (\mathbf{f}_t - \mathbf{P}_t^{(k)} \mathbf{f}_{ta}) \right\|^{2G} \\
&\leq C_9 \delta^{-2G} \xi_n^{-2G} \sum_{j=1}^K \sum_{k=1}^K \bar{\lambda}(\mathbf{P}_{\bar{c}}^{(j)'} \mathbf{P}_{\bar{c}}^{(j)}) \left\| \mathbf{A}_t(\mathbf{w}_k^0) \right\|^{2G} \\
&\leq C'_9 \delta^{-2G} \xi_n^{-2G} K \sum_{k=1}^K (R_{n_{\bar{c}}}(\mathbf{w}_k^0))^G \rightarrow 0.
\end{aligned}$$

Similarly, we have

$$\begin{aligned}
&P \left\{ \sup_{\mathbf{w} \in H_n} |\langle \mathbf{A}_t(\mathbf{w}), \mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta} \rangle| / R_n(\mathbf{w}) > \delta \right\} \\
&\leq C_{10} \delta^{-2G} \xi_n^{-2G} \sum_{j=1}^K \sum_{k=1}^K \bar{\lambda}(\mathbf{P}_t^{(j)'} \mathbf{P}_t^{(j)}) \left\| \mathbf{A}_t(\mathbf{w}_k^0) \mathbf{f}_t \right\|^{2G} \\
&\leq C'_{10} \delta^{-2G} \xi_n^{-2G} K \sum_{k=1}^K (R_{n_t}(\mathbf{w}_k^0))^G \rightarrow 0,
\end{aligned}$$

$$\begin{aligned}
& P \left\{ \sup_{\mathbf{w} \in H_n} |\langle \mathbf{A}_{\bar{c}}(\mathbf{w}), \mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta} \rangle| / R_n(\mathbf{w}) > \delta \right\} \\
& \leq C_{11} \delta^{-2G} \xi_n^{-2G} \sum_{j=1}^K \sum_{k=1}^K \bar{\lambda}(\mathbf{P}_t^{(j)'} \mathbf{P}_t^{(j)}) \|\mathbf{A}_{\bar{c}}(\mathbf{w}_k^0)\|^{2G} \\
& \leq C'_{11} \delta^{-2G} \xi_n^{-2G} K \sum_{k=1}^K (R_{\bar{c}}(\mathbf{w}_k^0))^G \rightarrow 0,
\end{aligned}$$

$$\begin{aligned}
& P \left\{ \sup_{\mathbf{w} \in H_n} |\langle \mathbf{A}_{\bar{c}}(\mathbf{w}), \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca} \rangle| / R_n(\mathbf{w}) > \delta \right\} \\
& \leq C_{12} \delta^{-2G} \xi_n^{-2G} \sum_{j=1}^K \sum_{k=1}^K \bar{\lambda}(\mathbf{P}_{\bar{c}}^{(j)'} \mathbf{P}_{\bar{c}}^{(j)}) \|\mathbf{A}_{\bar{c}}(\mathbf{w}_k^0)\|^{2G} \\
& \leq C'_{12} \delta^{-2G} \xi_n^{-2G} K \sum_{k=1}^K (R_{n_{\bar{c}}}(\mathbf{w}_k^0))^G \rightarrow 0
\end{aligned}$$

and

$$\begin{aligned}
& P \left\{ \sup_{\mathbf{w} \in H_n} |\langle \mathbf{P}_t(\mathbf{w}) \boldsymbol{\zeta}_{ta}, \mathbf{P}_{\bar{c}}(\mathbf{w}) \mathbf{v}_{ca} \rangle| / R_n(\mathbf{w}) > \delta \right\} \\
& \leq P \left\{ \sup_{\mathbf{w} \in H_n} \sum_{j=1}^K \sum_{k=1}^K w_j w_k |\boldsymbol{\zeta}'_{ta} \mathbf{P}_t^{(j)'} \mathbf{P}_{\bar{c}}^{(k)} \mathbf{v}_{ca}| > \delta \xi_n \right\} \\
& \leq P \left\{ \max_{1 \leq j \leq K} \max_{1 \leq k \leq K} |\boldsymbol{\zeta}'_{ta} \mathbf{P}_t^{(j)'} \mathbf{P}_{\bar{c}}^{(k)} \mathbf{v}_{ca}| > \delta \xi_n \right\} \\
& \leq \sum_{j=1}^K \sum_{k=1}^K P \left\{ |\boldsymbol{\zeta}'_{ta} \mathbf{P}_t^{(j)'} \mathbf{P}_{\bar{c}}^{(k)} \mathbf{v}_{ca}| > \delta \xi_n \right\} \\
& \leq \sum_{j=1}^K \sum_{k=1}^K \frac{E |\boldsymbol{\zeta}'_{ta} \mathbf{P}_t^{(j)'} \mathbf{P}_{\bar{c}}^{(k)} \mathbf{v}_{ca}|^2}{\delta^2 \xi_n^2} \\
& \leq \sum_{j=1}^K \sum_{k=1}^K \frac{E \|\mathbf{P}_t^{(j)} \boldsymbol{\zeta}_{ta}\|^2 E \|\mathbf{P}_{\bar{c}}^{(k)} \mathbf{v}_{ca}\|^2}{\delta^2 \xi_n^2} \\
& \leq C_{13} \delta^{-2} \xi_n^{-2} \sum_{j=1}^K \sum_{k=1}^K \text{tr} |\mathbf{P}_t(\mathbf{w}_j^0) \mathbf{P}_{\bar{c}}(\mathbf{w}_k^0)|^2 \\
& \leq C_{13} \delta^{-2} \xi_n^{-2} \bar{\lambda}(\mathbf{P}_t^2(\mathbf{w}^*)) \sum_{j=1}^K \sum_{k=1}^K [\text{tr}(\mathbf{P}_{\bar{c}}'(\mathbf{w}_j^0) \mathbf{P}_{\bar{c}}(\mathbf{w}_j^0))] \\
& \leq C'_{13} \delta^{-2} \xi_n^{-2} K \sum_{k=1}^K (R_{n_{\bar{c}}}(\mathbf{w}_k^0)) \rightarrow 0.
\end{aligned}$$

By the above results, it is readily seen that the right hand side of (A.26) converges to zero in probability. From (A.26) and noting that

$$\begin{aligned}
& \sup_{\mathbf{w} \in H_n} \left| \tilde{B}(\mathbf{w}) \right| / R_n(\mathbf{w}) \\
&= \sup_{\mathbf{w} \in H_n} \left| \underline{B}(\mathbf{w}) - L_n(\mathbf{w}) \right| / R_n(\mathbf{w}) \\
&= \sup_{\mathbf{w} \in H_n} \left[\left\| \underline{\Delta} + \mathbf{b} - \hat{\Delta}(\mathbf{w}) \right\|^2 - \left\| \underline{\Delta} - \hat{\Delta}(\mathbf{w}) \right\|^2 \right] / R_n(\mathbf{w}) \\
&\leq \sup_{\mathbf{w} \in H_n} \left[2(\underline{B}(\mathbf{w}) \|\mathbf{b}\|^2)^{1/2} \right] / R_n(\mathbf{w}) + \sup_{\mathbf{w} \in H_n} \|\mathbf{b}\|^2 / R_n(\mathbf{w}) \rightarrow_p 0, \quad (\text{A.27})
\end{aligned}$$

we can see that

$$\begin{aligned}
& \sup_{\mathbf{w} \in H_n} |L_n(\mathbf{w}) / R_n(\mathbf{w}) - 1| \\
&= \sup_{\mathbf{w} \in H_n} |[\{\underline{B}(\mathbf{w}) - L_n(\mathbf{w})\} - \{\underline{B}(\mathbf{w}) - R_n(\mathbf{w})\}]| / R_n(\mathbf{w}) \\
&= \sup_{\mathbf{w} \in H_n} |\underline{B}(\mathbf{w}) - L_n(\mathbf{w})| / R_n(\mathbf{w}) + \sup_{\mathbf{w} \in H_n} |\underline{B}(\mathbf{w}) - R_n(\mathbf{w})| / R_n(\mathbf{w}) \rightarrow_p 0,
\end{aligned}$$

which is (A.13). Thus, (A.3), (A.12) and (A.13) are proved, and this completes the proof of Theorem 2. \square

A.3 Proof of Theorem 3

Let

$$\begin{aligned}
\hat{C}_n(\mathbf{w}) &= \left\| \tilde{\mathbf{Y}} - \hat{\Delta}(\mathbf{w}) \right\|^2 + 2tr(\mathbf{P}_{t0}(\mathbf{w}))\hat{\sigma}_{K_t^*}^2 + 2tr(\mathbf{P}_{c0}(\mathbf{w}))\hat{\sigma}_{K_c^*}^2 \\
&= C_n(\mathbf{w}) + 2tr(\mathbf{P}_{t0}(\mathbf{w}))(\hat{\sigma}_{K_t^*}^2 - \sigma_t^2) + 2tr(\mathbf{P}_{c0}(\mathbf{w}))(\hat{\sigma}_{K_c^*}^2 - \sigma_c^2). \quad (\text{A.28})
\end{aligned}$$

By (A.28) and Theorem 2, Theorem 3 is valid if the following conditions hold

$$\sup_{\mathbf{w} \in H_n} tr(\mathbf{P}_{t0}(\mathbf{w})) \left| \hat{\sigma}_{K_t^*}^2 - \sigma_t^2 \right| / R_n(\mathbf{w}) \rightarrow_p 0 \quad (\text{A.29})$$

and

$$\sup_{\mathbf{w} \in H_n} tr(\mathbf{P}_{c0}(\mathbf{w})) \left| \hat{\sigma}_{K_c^*}^2 - \sigma_c^2 \right| / R_n(\mathbf{w}) \rightarrow_p 0. \quad (\text{A.30})$$

We need only to verify (A.30), (A.29) can be proved in the same way.

It should be mentioned that our assumption of \mathbf{U} being non-stochastic instead of allowing it to be stochastic has no impact on our proof, because all of our technical conditions related to \mathbf{U} hold almost surely. Let $r_{\bar{c}} = tr(\mathbf{P}_{c0}(\mathbf{w}))$. Then we have

$$r_{\bar{c}} \leq \bar{\lambda}(\mathbf{P}_{c0}(\mathbf{w})) l_{K^*} = (1 + o_p(1))l_{K^*} \quad (\text{A.31})$$

and

$$\begin{aligned}
& \sup_{\mathbf{w} \in H_n} \text{tr}(\mathbf{P}_{\tilde{c}0}(\mathbf{w})) \left| \widehat{\sigma}_{K^*}^2 - \sigma_c^2 \right| / R_n(\mathbf{w}) \\
& \leq \frac{r_{\tilde{c}}}{\xi_n} \left| \widehat{\sigma}_{K^*}^2 - \sigma_c^2 \right| = \frac{r_{\tilde{c}}}{\xi_n} \left| \frac{\mathbf{Y}'_{ca} (\mathbf{I} - \mathbf{P}_{ca}^{(K^*)}) \mathbf{Y}_{ca}}{n_c - l_{K^*}} - \sigma_c^2 \right| \\
& \leq \frac{r_{\tilde{c}}}{n_c - l_{K^*}} \frac{\mathbf{f}'_{ca} (\mathbf{I} - \mathbf{P}_{ca}^{(K^*)}) \mathbf{f}_{ca}}{\xi_n} + \frac{2r_{\tilde{c}} \left| \mathbf{f}'_{ca} (\mathbf{I} - \mathbf{P}_{ca}^{(K^*)}) \mathbf{v}_{ca} \right|}{\xi_n (n_c - l_{K^*})} \\
& \quad + \frac{r_{\tilde{c}} \left| \mathbf{v}'_{ca} (\mathbf{I} - \mathbf{P}_{ca}^{(K^*)}) \mathbf{v}_{ca} - (n_c - l_{K^*}) \sigma_c^2 \right|}{\xi_n (n_c - l_{K^*})}. \tag{A.32}
\end{aligned}$$

It suffices to show that the three terms on the right hand of (A.32) tend to zero in probability. Using Condition (25), we obtain

$$\frac{\mathbf{f}'_{ca} (\mathbf{I} - \mathbf{P}_{ca}^{(K^*)}) \mathbf{f}_{ca}}{\xi_n^2} \rightarrow 0. \tag{A.33}$$

Then from (24), (26), (A.31) and (A.33), it is seen that

$$\begin{aligned}
& \frac{r_{\tilde{c}}}{n_c - l_{K^*}} \frac{\mathbf{f}'_{ca} (\mathbf{I} - \mathbf{P}_{ca}^{(K^*)}) \mathbf{f}_{ca}}{\xi_n} \\
& \leq \left[\frac{r_{\tilde{c}}^2}{n_c - l_{K^*}} \frac{\mathbf{f}'_{ca} (\mathbf{I} - \mathbf{P}_{ca}^{(K^*)}) \mathbf{f}_{ca}}{\xi_n^2} \frac{\mathbf{f}'_{ca} \mathbf{f}_{ca}}{n_c - l_{K^*}} \right]^{1/2} \rightarrow 0 \tag{A.34}
\end{aligned}$$

and

$$\begin{aligned}
& P \left\{ \frac{2r_{\tilde{c}} \left| \mathbf{f}'_{ca} (\mathbf{I} - \mathbf{P}_{ca}^{(K^*)}) \mathbf{v}_{ca} \right|}{\xi_n (n_c - l_{K^*})} > \delta \right\} \\
& \leq E \left[\mathbf{f}'_{ca} (\mathbf{I} - \mathbf{P}_{ca}^{(K^*)}) \mathbf{v}_{ca} \right]^2 \frac{4r_{\tilde{c}}^2}{\delta^2 \xi_n^2 (n_c - l_{K^*})^2} \\
& \leq \frac{C_{14} r_{\tilde{c}}^2 \mathbf{f}'_{ca} (\mathbf{I} - \mathbf{P}_{ca}^{(K^*)}) \mathbf{f}_{ca}}{\delta^2 \xi_n^2 (n_c - l_{K^*})^2} \rightarrow 0. \tag{A.35}
\end{aligned}$$

By (A.35), it can be seen the second term on the right hand side of (A.32) tends to zero in probability. From

$$\begin{aligned}
& P \left\{ \frac{r_{\tilde{c}} \left| \mathbf{v}'_{ca} (\mathbf{I} - \mathbf{P}_{ca}^{(K^*)}) \mathbf{v}_{ca} - (n_c - l_{K^*}) \sigma_c^2 \right|}{\xi_n (n_c - l_{K^*})} > \delta \right\} \\
& \leq E \left[\mathbf{v}'_{ca} (\mathbf{I} - \mathbf{P}_{ca}^{(K^*)}) \mathbf{v}_{ca} - (n_c - l_{K^*}) \sigma_c^2 \right]^2 \frac{r_{\tilde{c}}^2}{\delta^2 \xi_n^2 (n_c - l_{K^*})^2} \\
& \leq \frac{C_{15} r_{\tilde{c}}^2 (n_c - l_{K^*})}{\delta^2 \xi_n^2 (n_c - l_{K^*})^2} \rightarrow 0, \tag{A.36}
\end{aligned}$$

we see the third term on the right hand side of (A.32) tends to zero in probability. Combining (A.32)–(A.36), it is straightforward to see that (A.30) holds. This completes the proof of Theorem 3. \square

A.4 An example in which Condition (14) is satisfied

Consider the data-generating process

$$Y_i = [f_t(z_i) + \zeta_i] I(T_i = t) + [f_c(z_i) + v_i] I(T_i = c), \quad 1 \leq i \leq n,$$

where $f_t(z_i) = -f_c(z_i) = \sum_{s=1}^{\infty} \theta_s \cos((s-1)z_i)/s$, $z_i = 2\pi(i-1)/n$, and ζ_i and v_i are i.i.d. $N(0, \sigma^2)$. Suppose that $\theta_s = s^{-7/12}$, and we smooth across estimators from models that comprise subsets of the first p ($p < S$) regressors, where S is the largest integer that is no greater than $n/2$. Altogether there are $K = 2^{p-1}$ approximating models if we assume all models contain the intercept.

From the proof of Example 1 in Wan et al. (2010), it can be seen that $\frac{np^{-13/6}}{\xi_n} = o(1)$. Hence Condition (14) holds if $p = O((\log n)^{1/5})$.

A.5 Results of Design 2

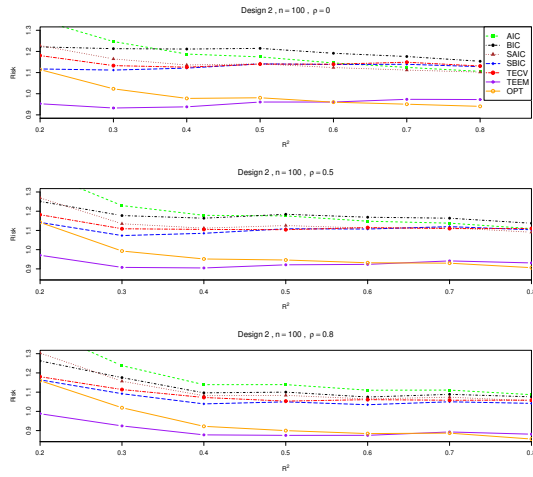
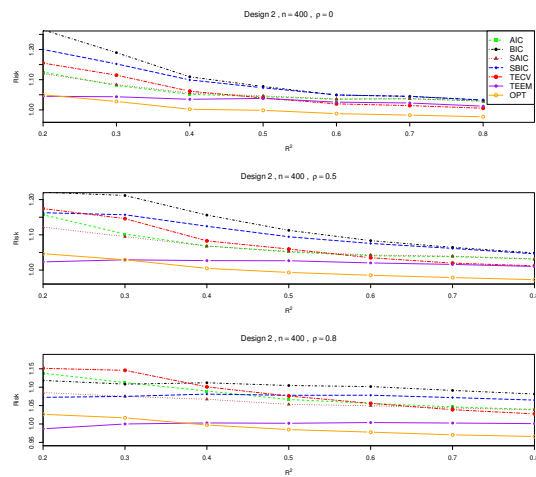
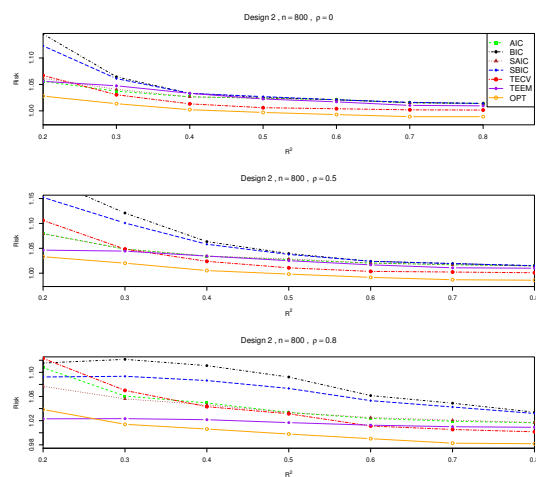


Fig. 6: Results of simulations under Design 2 with $n = 100$.

Fig. 7: Results of simulations under Design 2 with $n = 400$.Fig. 8: Results of simulations under Design 2 with $n = 800$.

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