

A Proofs

A.1 Proof of Theorem 1

1. Consistency of $\hat{\boldsymbol{\lambda}}$.

Let us define

$$\begin{aligned}\boldsymbol{\psi}_n(\boldsymbol{\lambda}) &= \frac{1}{N} \left\{ \sum_{k \in S} d_k R_k F(\boldsymbol{\lambda}^\top \mathbf{Z}_k) \mathbf{X}_k - \sum_{k \in P} \mathbf{X}_k \right\}, \\ \boldsymbol{\psi}(\boldsymbol{\lambda}) &= \mathbb{E} \left\{ (R_k F(\boldsymbol{\lambda}^\top \mathbf{Z}_k) - 1) \mathbf{X}_k \right\}.\end{aligned}$$

Let $\Psi_n(\boldsymbol{\lambda}) = -\|\boldsymbol{\psi}_n(\boldsymbol{\lambda})\|$ and $\Psi(\boldsymbol{\lambda}) = -\|\boldsymbol{\psi}(\boldsymbol{\lambda})\|$. Hence, $\hat{\boldsymbol{\lambda}} \in \arg \max_{\boldsymbol{\lambda} \in \boldsymbol{\Lambda}} \Psi_n(\boldsymbol{\lambda})$. We check the conditions of Theorem 5.9 in van der Vaart (2000). First,

$$\mathbb{E}(\boldsymbol{\psi}_n(\boldsymbol{\lambda})|\mathcal{U}) = \frac{1}{N} \sum_{k \in P} [R_k F(\boldsymbol{\lambda}^\top \mathbf{Z}_k) - 1] \mathbf{X}_k,$$

which implies that $\mathbb{E}\{\boldsymbol{\psi}_n(\boldsymbol{\lambda})\} = \boldsymbol{\psi}(\boldsymbol{\lambda})$. Since $R_k F(\boldsymbol{\lambda}^\top \mathbf{Z}_k) - 1$ is bounded, it follows that $\{R_k F(\boldsymbol{\lambda}^\top \mathbf{Z}_k) - 1\} \mathbf{X}_k$ admits second-order moments. Then, $\mathbb{V}[\mathbb{E}(\boldsymbol{\psi}_n(\boldsymbol{\lambda})|\mathcal{U})]$ converges to 0. Also,

$$\mathbb{E}[\mathbb{V}\{\boldsymbol{\psi}_n(\boldsymbol{\lambda})|\mathcal{U}\}] = \mathbb{E} \left[\mathbb{V} \left\{ \frac{1}{N} \sum_{k \in S} d_k R_k F(\boldsymbol{\lambda}^\top \mathbf{Z}_k) \mathbf{X}_k \middle| \mathcal{U} \right\} \right]. \quad (25)$$

By Assumption 3-(i), the right-hand side of (25) tends to 0. Then $\boldsymbol{\psi}_n(\boldsymbol{\lambda}) \rightarrow \boldsymbol{\psi}(\boldsymbol{\lambda})$ in L^2 , and thus in probability.

Because \mathbf{Z}_k has compact support and $\boldsymbol{\Lambda}$ is compact, there exists a compact interval I including with probability one the interval $[\min_{\boldsymbol{\lambda} \in \boldsymbol{\Lambda}} \boldsymbol{\lambda}^\top \mathbf{Z}_k, \max_{\boldsymbol{\lambda} \in \boldsymbol{\Lambda}} \boldsymbol{\lambda}^\top \mathbf{Z}_k]$. Moreover, $F(\cdot)$ is uniformly continuous on I . Now, fix ε and let δ be such that for any $(a, b) \in I^2$, $|a - b| < \delta$ implies $|F(a) - F(b)| < \varepsilon$. Let C be such that $\|\mathbf{Z}_k\| \leq C$ with probability one. Consider balls of center $\boldsymbol{\lambda}_b$ and of radius δ/C for $\boldsymbol{\lambda}$. Then, for $\boldsymbol{\lambda}$ within such a ball, we get, by the triangular and Cauchy-Schwarz inequalities,

$$\begin{aligned}\|\boldsymbol{\psi}_n(\boldsymbol{\lambda}) - \boldsymbol{\psi}_n(\boldsymbol{\lambda}_b)\| &\leq \frac{1}{N} \sum_{k \in S} d_k R_k |F(\boldsymbol{\lambda}^\top \mathbf{Z}_k) - F(\boldsymbol{\lambda}_b^\top \mathbf{Z}_k)| \|\mathbf{X}_k\| \\ &\leq \frac{\varepsilon}{N} \sum_{k \in S} d_k R_k \|\mathbf{X}_k\|.\end{aligned} \quad (26)$$

Similarly, $\|\psi(\boldsymbol{\lambda}) - \psi(\boldsymbol{\lambda}_b)\| \leq \varepsilon \mathbb{E}(\|\mathbf{X}_k\|)$. Now, by assumption, $\boldsymbol{\Lambda}$ is compact. It can then be recovered by B balls of centers $\boldsymbol{\lambda}_b$ ($b = 1 \dots B$) and of radius δ/C . Then, using $|||a| - |b||| \leq ||a - b||$, we get

$$\begin{aligned} \sup_{\boldsymbol{\lambda} \in \boldsymbol{\Lambda}} |\Psi_n(\boldsymbol{\lambda}) - \Psi(\boldsymbol{\lambda})| &\leq \sup_{\boldsymbol{\lambda} \in \boldsymbol{\Lambda}} \|\psi_n(\boldsymbol{\lambda}) - \psi(\boldsymbol{\lambda})\| \\ &\leq \max_{b=1 \dots B} \|\psi_n(\boldsymbol{\lambda}_b) - \psi(\boldsymbol{\lambda}_b)\| + \varepsilon \left\{ \mathbb{E}(\|\mathbf{X}_k\|) + \frac{1}{N} \sum_{k \in S} d_k R_k \|\mathbf{X}_k\| \right\}. \end{aligned}$$

The first term on the right-hand side converges in probability to zero by pointwise consistency of $\psi_n(\boldsymbol{\lambda})$. By Assumption 3-(i) and reasoning as above, the term within brackets converges in probability to $\mathbb{E}\{(1 + R_k)\|\mathbf{X}_k\|\}$. Therefore, with probability tending to one, the left-hand side is smaller than $\varepsilon[1 + 2\mathbb{E}\{(1 + R_k)\|\mathbf{X}_k\|\}]$. Because ε was arbitrary, we have proved

$$\sup_{\boldsymbol{\lambda} \in \boldsymbol{\Lambda}} |\Psi_n(\boldsymbol{\lambda}) - \Psi(\boldsymbol{\lambda})| \xrightarrow{P} 0.$$

Hence, condition (i) in Theorem 5.9 of van der Vaart (2000) holds.

We now check condition (ii). First, by Assumptions 1 and 2,

$$\mathbb{E}[(R_k F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) - 1) \mathbf{X}_k | \mathbf{Z}_k] = [F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) \mathbb{E}(R_k | \mathbf{Z}_k) - 1] \mathbb{E}(\mathbf{X}_k | \mathbf{Z}_k) = 0.$$

Thus, $\psi(\boldsymbol{\lambda}_0) = 0$ and $\Psi(\boldsymbol{\lambda}_0) = 0$. Suppose that there exists $\boldsymbol{\lambda}_1$ such that $\Psi(\boldsymbol{\lambda}_1) = 0$. Then $\psi(\boldsymbol{\lambda}_1) = 0$ and because $\mathbb{E}(\mathbf{X}_k | R_k = 1, \mathbf{Z}_k) = \boldsymbol{\Gamma} \mathbf{Z}_k$, we obtain, by the law of iterated expectation,

$$\boldsymbol{\Gamma} \mathbb{E}[R_k \{F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) - F(\boldsymbol{\lambda}_1^\top \mathbf{Z}_k)\} \mathbf{Z}_k] = 0.$$

Because the rank of $\boldsymbol{\Gamma}$ is equal to $\dim(\mathbf{Z}_k)$,

$$\mathbb{E}[R_k \{F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) - F(\boldsymbol{\lambda}_1^\top \mathbf{Z}_k)\} \mathbf{Z}_k] = 0.$$

This, in turn, implies that

$$\mathbb{E}[R_k (F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) - F(\boldsymbol{\lambda}_1^\top \mathbf{Z}_k)) (\boldsymbol{\lambda}_0^\top \mathbf{Z}_k - \boldsymbol{\lambda}_1^\top \mathbf{Z}_k)] = 0. \quad (27)$$

Now, because $F(\cdot)$ is strictly increasing, we have

$$\{F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) - F(\boldsymbol{\lambda}_1^\top \mathbf{Z}_k)\} (\boldsymbol{\lambda}_0^\top \mathbf{Z}_k - \boldsymbol{\lambda}_1^\top \mathbf{Z}_k) \geq 0 \text{ with equality iff } \boldsymbol{\lambda}_0^\top \mathbf{Z}_k - \boldsymbol{\lambda}_1^\top \mathbf{Z}_k = 0.$$

Hence, (27) implies that $(\boldsymbol{\lambda}_0 - \boldsymbol{\lambda}_1)^\top \mathbf{Z}_k = 0$ almost surely. This and the fact that $\mathbb{E}(\mathbf{Z}_k \mathbf{Z}_k^\top)$ is nonsingular imply that $\boldsymbol{\lambda}_1 = \boldsymbol{\lambda}_0$.

Thus, $\Psi(\boldsymbol{\lambda}) = 0$ implies that $\boldsymbol{\lambda} = \boldsymbol{\lambda}_0$. Second, by the same argument following (26), we have, for any $\boldsymbol{\lambda}, \boldsymbol{\lambda}'$ such that $\|\boldsymbol{\lambda} - \boldsymbol{\lambda}'\| < \delta/C$,

$$\|\Psi(\boldsymbol{\lambda}) - \Psi(\boldsymbol{\lambda}')\| \leq \|\boldsymbol{\psi}(\boldsymbol{\lambda}) - \boldsymbol{\psi}(\boldsymbol{\lambda}')\| \leq \varepsilon \mathbb{E}(\|\mathbf{X}_k\|).$$

Hence, Ψ is continuous. Thus, for any $\varepsilon' > 0$,

$$\inf_{\boldsymbol{\lambda} \in \boldsymbol{\Lambda}: \|\boldsymbol{\lambda} - \boldsymbol{\lambda}_0\| \geq \varepsilon} \|\Psi(\boldsymbol{\lambda})\| = \min_{\boldsymbol{\lambda} \in \boldsymbol{\Lambda}: \|\boldsymbol{\lambda} - \boldsymbol{\lambda}_0\| \geq \varepsilon} \|\Psi(\boldsymbol{\lambda})\| > 0 = \|\Psi(\boldsymbol{\lambda}_0)\|$$

and condition (ii) in Theorem 5.9 of van der Vaart (2000) holds.

As a result, both conditions of this theorem are satisfied, and $\widehat{\boldsymbol{\lambda}}$ is consistent.

Consistency of \widehat{t}_C .

First,

$$\begin{aligned} (\widehat{t}_C - t_y) / N &= \frac{1}{N} \sum_{k \in S} d_k R_k F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) Y_k - \frac{1}{N} \sum_{k \in P} Y_k \\ &\quad + \frac{1}{N} \sum_{k \in S} d_k R_k \left\{ F(\widehat{\boldsymbol{\lambda}}^\top \mathbf{Z}_k) - F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) \right\} Y_k. \end{aligned} \quad (28)$$

We now show that both terms on the right-hand side of (28), denoted by A_1 and A_2 hereafter, tend to zero in probability. For A_1 , we use arguments similar to those used for the pointwise consistency of $\boldsymbol{\psi}_n(\lambda)$. We have

$$\mathbb{E}(A_1 | \mathcal{U}) = \frac{1}{N} \sum_{k \in P} \{ R_k F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) - 1 \} Y_k.$$

Moreover, by the law of iterated expectation and (10),

$$\mathbb{E} \{ [R_k F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) - 1] Y_k \} = \mathbb{E} \{ [\mathbb{E}(R_k | \mathbf{Z}_k) F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) - 1] \mathbb{E}[Y_k | \mathbf{Z}_k] \} = 0.$$

Hence, $\mathbb{E}(A_1) = 0$ and $\mathbb{V}\{\mathbb{E}(A_1 | \mathcal{U})\}$ converges to 0. Moreover,

$$\mathbb{E} \{ \mathbb{V}(A_1 | \mathcal{U}) \} = \mathbb{E} \left\{ \mathbb{V} \left(\frac{1}{N} \sum_{k \in S} d_k R_k F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) Y_k \middle| \mathcal{U} \right) \right\},$$

and the right-hand side converges to 0 by Assumption 3-(i). Thus, $\mathbb{V}(A_1)$ tends to 0 and A_1 converges to 0 in probability.

Turning to A_2 , note that $F(\cdot)$ is uniformly continuous on the compact set. Set $\varepsilon > 0$ and δ as above. By consistency of $\hat{\boldsymbol{\lambda}}$, $\|\hat{\boldsymbol{\lambda}} - \boldsymbol{\lambda}_0\| < \delta/C$ with probability close to one. Then, by the Cauchy-Schwarz inequality, $\max_k \|(\hat{\boldsymbol{\lambda}} - \boldsymbol{\lambda}_0)^\top \mathbf{Z}_k\| < \delta$, implying that

$$\max_k \left| F(\hat{\boldsymbol{\lambda}}^\top \mathbf{Z}_k) - F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) \right| < \varepsilon,$$

with a probability close to one. Hence, with such a probability,

$$|A_2| < \varepsilon \left(\frac{1}{N} \sum_{k \in S} d_k |Y_k| \right).$$

By Assumption 3-(i), the term into parentheses converges to $\mathbb{E}(|Y_k|)$. Therefore, A_2 converges to zero, and the result follows. \square

2. Linearization of $\hat{\boldsymbol{\lambda}} - \boldsymbol{\lambda}_0$.

Let us define $\hat{\mathbf{G}}(\boldsymbol{\lambda}) = \frac{1}{N} \sum_{k \in S} d_k R_k F'(\boldsymbol{\lambda}^\top \mathbf{Z}_k) \mathbf{X}_k \mathbf{Z}_k^\top$. Then, by the first-order condition of (5) and the mean value theorem,

$$\begin{aligned} 0 &= \frac{\hat{\mathbf{G}}(\hat{\boldsymbol{\lambda}})^\top}{N} \left\{ \sum_{k \in S} d_k R_k F(\hat{\boldsymbol{\lambda}}^\top \mathbf{Z}_k) \mathbf{X}_k - \sum_{k \in P} \mathbf{X}_k \right\} \\ &= \frac{\hat{\mathbf{G}}(\hat{\boldsymbol{\lambda}})^\top}{N} \left\{ \sum_{k \in S} d_k R_k F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) \mathbf{X}_k - \sum_{k \in P} \mathbf{X}_k \right\} + \hat{\mathbf{G}}(\hat{\boldsymbol{\lambda}})^\top \hat{\mathbf{G}}(\tilde{\boldsymbol{\lambda}})(\hat{\boldsymbol{\lambda}} - \boldsymbol{\lambda}_0), \end{aligned} \quad (29)$$

where $\tilde{\boldsymbol{\lambda}} = \tilde{t}\boldsymbol{\lambda}_0 + (1 - \tilde{t})\hat{\boldsymbol{\lambda}}$ for some $\tilde{t} \in [0, 1]$.

Because $F'(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k)$ is bounded, we have, by the same arguments as when showing $\boldsymbol{\psi}_n(\boldsymbol{\lambda}) \xrightarrow{P} \boldsymbol{\psi}(\boldsymbol{\lambda})$,

$$\hat{\mathbf{G}}(\boldsymbol{\lambda}_0) \xrightarrow{P} \mathbf{G}.$$

Now fix $\varepsilon > 0$. F' is continuous, and therefore uniformly continuous on the interval I defined above. Thus, there exists $\delta_2 > 0$ such that for any $(a, b) \in I^2$, $|a - b| < \delta_2$ implies $|F'(a) - F'(b)| < \varepsilon$. By Step 1, with a large probability, $\|\hat{\boldsymbol{\lambda}} - \boldsymbol{\lambda}_0\| < \delta_2/C$

and $\|\tilde{\boldsymbol{\lambda}} - \boldsymbol{\lambda}_0\| < \delta_2/C$. Then, by the triangular and Cauchy-Schwarz inequality, with a large probability,

$$\left\| \hat{\mathbf{G}}(\hat{\boldsymbol{\lambda}}) - \hat{\mathbf{G}}(\boldsymbol{\lambda}_0) \right\| \leq \frac{\varepsilon}{N} \sum_{k \in S} d_k R_k \|\mathbf{X}_k \mathbf{Z}_k^\top\|.$$

The same holds if $\hat{\boldsymbol{\lambda}}$ is replaced with $\tilde{\boldsymbol{\lambda}}$. Because \mathbf{Z}_k is bounded, the right-hand side converges to 0 by Assumption 3-(i). Hence, both $\hat{\mathbf{G}}(\hat{\boldsymbol{\lambda}})$ and $\hat{\mathbf{G}}(\tilde{\boldsymbol{\lambda}})$ converge in probability to \mathbf{G} . This and (29) imply that

$$\hat{\boldsymbol{\lambda}} - \boldsymbol{\lambda}_0 = -\frac{(\mathbf{G}^\top \mathbf{G})^{-1} \mathbf{G}^\top}{N} \left\{ \sum_{k \in S} d_k R_k F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) \mathbf{X}_k - \sum_{k \in P} \mathbf{X}_k \right\} \{1 + o_P(1)\}.$$

Asymptotic normality of \hat{t}_C .

First, we have

$$\begin{aligned} \frac{\hat{t}_C}{N} &= \frac{1}{N} \sum_{k \in S} d_k R_k F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) Y_k + \frac{1}{N} \sum_{k \in S} d_k R_k \left\{ F(\hat{\boldsymbol{\lambda}}^\top \mathbf{Z}_k) - F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) \right\} Y_k \\ &= \frac{1}{N} \sum_{k \in S} d_k R_k F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) Y_k - \left\{ \frac{1}{N} \sum_{k \in S} d_k R_k Y_k F'(\bar{\boldsymbol{\lambda}}^\top \mathbf{Z}_k) \mathbf{Z}_k^\top \right\} \\ &\quad \times \frac{(\mathbf{G}^\top \mathbf{G})^{-1} \mathbf{G}^\top}{N} \left\{ \sum_{k \in S} d_k R_k F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) \mathbf{X}_k - \sum_{k \in P} \mathbf{X}_k \right\} \{1 + o_P(1)\}, \end{aligned}$$

where $\bar{\boldsymbol{\lambda}} = \bar{t} \boldsymbol{\lambda}_0 + (1 - \bar{t}) \hat{\boldsymbol{\lambda}}$ for some $\bar{t} \in [0, 1]$. By the same argument as above,

$$\frac{1}{N} \sum_{k \in S} d_k R_k Y_k F'(\bar{\boldsymbol{\lambda}}^\top \mathbf{Z}_k) \mathbf{Z}_k^\top \xrightarrow{P} \mathbb{E}(\rho_k Y_k \mathbf{Z}_k^\top).$$

Hence, since $\boldsymbol{\gamma} = \mathbf{G}(\mathbf{G}^\top \mathbf{G})^{-1} \mathbb{E}(\rho_k Y_k \mathbf{Z}_k^\top)$ and $W_k = (R_k F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) - 1)(Y_k - \boldsymbol{\gamma}^\top \mathbf{X}_k)$, we get

$$\hat{t}_C - t_y = \left\{ \sum_{k \in S} d_k (W_k + Y_k - \boldsymbol{\gamma}^\top \mathbf{X}_k) - \sum_{k \in P} W_k + Y_k - \boldsymbol{\gamma}^\top \mathbf{X}_k + \sum_{k \in P} W_k \right\} \{1 + o_P(1)\}. \quad (30)$$

To prove the result, we now check the conditions of Theorem 2 in Chen and Rao (2007). Note first that \mathcal{U} plays the role of \mathcal{B}_n in their theorem, $\sum_{k \in S} d_k (W_k + Y_k - \boldsymbol{\gamma}^\top \mathbf{X}_k) - \sum_{k \in P} W_k + Y_k - \boldsymbol{\gamma}^\top \mathbf{X}_k$ corresponds to U_n , $\sum_{k \in P} W_k$ corresponds to V_n , $\sigma_{2N} \equiv$

$\mathbb{V} \left\{ \sum_{k \in S} d_k(W_k + Y_k - \gamma^\top \mathbf{X}_k) \middle| \mathcal{U} \right\}^{1/2}$ corresponds to σ_{2n} and $\sigma_{1N} \equiv \mathbb{V} \left(\sum_{k \in P} W_k \right)^{1/2}$ corresponds to σ_{1n} .

Then, by Assumptions 1 and 2,

$$\mathbb{E}(W_k | \mathbf{Z}_k) = [\mathbb{E}(R_k | \mathbf{Z}_k) F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) - 1] \mathbb{E}(Y_k - \gamma^\top \mathbf{X}_k | \mathbf{Z}_k) = 0.$$

Hence, $\mathbb{E}(W_k) = 0$ and by the central limit theorem,

$$\sigma_{1N}^{-1} \left(\sum_{k \in P} W_k \right) \xrightarrow{d} \mathcal{N}(0, 1).$$

Also, $\sum_{k \in P} W_k$ is \mathcal{U} -measurable. Hence, their condition 1 holds.

Next,

$$\mathbb{E} \left\{ \sum_{k \in S} d_k(W_k + Y_k - \gamma^\top \mathbf{X}_k) - \sum_{k \in P} W_k + Y_k - \gamma^\top \mathbf{X}_k \middle| \mathcal{U} \right\} = 0$$

and Condition (1) in Chen and Rao (2007) holds by Assumption 3-(iii) and Polya's theorem (see the remark below Theorem 2 in Chen and Rao, 2007). Thus, their condition 2 holds. Finally their condition 3 holds by Assumption 3-(ii).

Therefore,

$$\frac{\sum_{k \in S} d_k(W_k + Y_k - \gamma^\top \mathbf{X}_k) - \sum_{k \in P} W_k + Y_k - \gamma^\top \mathbf{X}_k + \sum_{k \in P} W_k}{\sqrt{\sigma_{1N}^2 + \sigma_{2N}^2}} \xrightarrow{d} \mathcal{N}(0, 1).$$

The result follows by (30) and Slutsky's lemma. \square

A.2 Proof of Theorem 2

1. By assumption, a solution to (14) exists. By the same reasoning as the one used to show $\boldsymbol{\lambda}_0 = \boldsymbol{\lambda}_1$ in the proof of Theorem 1, the solution is unique. Moreover, still reasoning as in the first step of the proof of Theorem 1, we have $\widehat{\boldsymbol{\lambda}} \xrightarrow{P} \boldsymbol{\lambda}_\infty$. \square
2. We can decompose the total error of estimation of \widehat{t}_C as in (13). Using Assumption 3 and the same arguments as in the second step of the proof of Theorem 1, the first three terms of the right-hand side converge to 0 in probability. On the other hand, the

fourth term on the right-hand side does not tend to zero in probability. By the law of large number and Assumption 3-(i), it converges towards

$$\mathbb{E} \left[R_k \left\{ F(\boldsymbol{\lambda}_\infty^\top \mathbf{Z}_k) - F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) \right\} Y_k \right].$$

This in turn can be rewritten as:

$$\mathbb{E} \left\{ R_k \left\{ F(\boldsymbol{\lambda}_\infty^\top \mathbf{Z}_k) - F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) \right\} Y_k \right\} = \mathbb{E} \left(f_k R_k Y_k \mathbf{Z}_k^\top \right) (\boldsymbol{\lambda}_\infty - \boldsymbol{\lambda}_0),$$

where f_k is defined in (16). Next, we prove (15). We have

$$\begin{aligned} -\mathbb{E} \left[f_k F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k)^{-1} X_k \mathbf{Z}_k^\top (\boldsymbol{\lambda}_\infty - \boldsymbol{\lambda}_0) \right] &= \mathbb{E} \left\{ \left(1 - \frac{F(\boldsymbol{\lambda}_\infty^\top \mathbf{Z}_k)}{F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k)} \right) X_k \right\} \\ &= \mathbb{E} \left\{ F(\boldsymbol{\lambda}_\infty^\top \mathbf{Z}_k) (R_k X_k - \mathbb{E}(R_k | \mathbf{Z}_k) X_k) \right\} \\ &= \mathbb{E} \left\{ F(\boldsymbol{\lambda}_\infty^\top \mathbf{Z}_k) \mathbb{C}ov(\mathbf{X}_k, R_k | \mathbf{Z}_k) \right\}, \end{aligned}$$

where the second equality comes from the nonresponse model and (14), and the third equality from the law of iterated expectation. This shows (15), which in turn implies that

$$\begin{aligned} &\mathbb{E} \left\{ R_k \left\{ F(\boldsymbol{\lambda}_\infty^\top \mathbf{Z}_k) - F(\boldsymbol{\lambda}_0^\top \mathbf{Z}_k) \right\} Y_k \right\} \\ &= - \left\{ \mathbb{E}(f_k R_k \mathbf{Z}_k \mathbf{X}_k^\top) \right\}^{-1} \mathbb{E}(f_k R_k \mathbf{Z}_k Y_k) \right\}^\top \mathbb{E} \left\{ F(\boldsymbol{\lambda}_\infty^\top \mathbf{Z}_k) \mathbb{C}ov(\mathbf{X}_k, R_k | \mathbf{Z}_k) \right\}. \square \end{aligned}$$