

# Supplementary material for “Score test for unconfoundedness under a logistic treatment assignment model”

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## Abstract

This supplementary materials consists of three sections. In Section 1, we review the conditions and Theorem 1 in the main paper. Section 2 contains a lemma, which can ease much burden of our proof of Theorem 1, which is given in Section 3.

## 1 Conditions and Theorem 1

We begin by reviewing the conditions we made and Theorem 1 in the main paper.

**Condition (C1)**  $\mathbb{E}\|\mathbf{X}\|^2 < \infty$  and  $\mathbf{A} = \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\mathbf{X}\mathbf{X}^\top]$  is of full rank.

**Condition (C2)** (i) The parameter space  $\Omega_1$  of  $\boldsymbol{\zeta}$  is independent of  $(y, \mathbf{x})$  and compact. (ii) The true value  $\boldsymbol{\zeta}_0$  of  $\boldsymbol{\zeta}$  is an interior point of  $\Omega_1$ . (iii)  $\boldsymbol{\zeta}$  is identifiable, i.e.  $\mathbb{E}\{\int |f\{y(1)|\mathbf{X}, \boldsymbol{\zeta}\} - f\{y(1)|\mathbf{X}, \boldsymbol{\zeta}'\}|dy\} > 0$  for any different elements  $\boldsymbol{\zeta}$  and  $\boldsymbol{\zeta}'$  in  $\Omega_1$ . (iv)  $\mathbb{E}\{\sup_{\boldsymbol{\zeta} \in \Omega_1} |\log f\{y(1)|\mathbf{X}, \boldsymbol{\zeta}\}|\} < \infty$ . (v)  $f\{y(1)|\mathbf{X}, \boldsymbol{\zeta}\}$  is continuous in  $\boldsymbol{\zeta}$  for almost all  $(y, \mathbf{x})$ .

**Condition (C3)** (i) The parameter space  $\Omega_2$  of  $\boldsymbol{\eta}$  is independent of  $(y, \mathbf{x})$  and compact. (ii) The true value  $\boldsymbol{\eta}_0$  of  $\boldsymbol{\eta}$  is an interior point of  $\Omega_2$ . (iii)  $\boldsymbol{\eta}$  is identifiable, i.e.  $\mathbb{E}\{\int |f\{y(0)|\mathbf{X}, \boldsymbol{\eta}\} - f\{y(0)|\mathbf{X}, \boldsymbol{\eta}'\}|dy\} > 0$  for any different elements  $\boldsymbol{\eta}$  and  $\boldsymbol{\eta}'$  in  $\Omega_2$ . (iv)  $\mathbb{E}\{\sup_{\boldsymbol{\eta} \in \Omega_2} |\log f\{y(0)|\mathbf{X}, \boldsymbol{\eta}\}|\} < \infty$ . (v)  $f\{y(0)|\mathbf{X}, \boldsymbol{\eta}\}$  is continuous in  $\boldsymbol{\eta}$  for almost all  $(y, \mathbf{x})$ .

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**Condition (C4)** (i)  $f\{y(1)|\mathbf{X}, \boldsymbol{\zeta}\}$  is twice differentiable with respect to  $\boldsymbol{\zeta}$  for almost all  $(y, \mathbf{x})$ , and  $\nabla_{\boldsymbol{\zeta}\boldsymbol{\zeta}^\top} f\{y(1)|\mathbf{X}, \boldsymbol{\zeta}\}$  is continuous at  $\boldsymbol{\zeta}_0$ . (ii)  $\mathbf{B} = \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{\nabla_{\boldsymbol{\zeta}} \log f\{y(1)|\mathbf{X}, \boldsymbol{\zeta}\}\}^{\otimes 2}]$  is positive definite. (iii) There exist a  $\delta > 0$  and positive functions  $M_1(\mathbf{x})$  and  $M_2(y, \mathbf{x})$  such that  $\mathbb{E}\{M_1(\mathbf{X})\} < \infty$  and  $\mathbb{E}\{M_2(Y, \mathbf{X})\} < \infty$ , and  $\|\mathbf{x}\| \int |t| \{f(t|\mathbf{x}, \boldsymbol{\zeta}) + \|\nabla_{\boldsymbol{\zeta}} f(t|\mathbf{x}, \boldsymbol{\zeta})\|\} dt \leq M_1(\mathbf{x})$  and  $\|\nabla_{\boldsymbol{\zeta}\boldsymbol{\zeta}^\top} \log f(y|\mathbf{x}, \boldsymbol{\zeta})\| \leq M_2(y, \mathbf{x})$  for all  $\boldsymbol{\zeta}$  satisfying  $\|\boldsymbol{\zeta} - \boldsymbol{\zeta}_0\| \leq \delta$ .

**Condition (C5)** (i)  $f\{y(0)|\mathbf{X}, \boldsymbol{\eta}\}$  is twice differentiable with respect to  $\boldsymbol{\eta}$  for almost all  $(y, \mathbf{x})$ , and  $\nabla_{\boldsymbol{\eta}\boldsymbol{\eta}^\top} f\{y(0)|\mathbf{X}, \boldsymbol{\eta}\}$  is continuous at  $\boldsymbol{\eta}_0$ . (ii)  $\mathbf{C} = \mathbb{E}[\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{\nabla_{\boldsymbol{\eta}} \log f\{y(0)|\mathbf{X}, \boldsymbol{\eta}\}\}^{\otimes 2}]$  is positive definite. (iii) There exist a  $\delta > 0$  and positive functions  $M_3(\mathbf{x})$  and  $M_4(y, \mathbf{x})$  such that  $\mathbb{E}\{M_3(\mathbf{X})\} < \infty$  and  $\mathbb{E}\{M_4(Y, \mathbf{X})\} < \infty$ , and  $\|\mathbf{x}\| \int |t| \{f(t|\mathbf{x}, \boldsymbol{\eta}) + \|\nabla_{\boldsymbol{\eta}} f(t|\mathbf{x}, \boldsymbol{\eta})\|\} dt \leq M_3(\mathbf{x})$  and  $\|\nabla_{\boldsymbol{\eta}\boldsymbol{\eta}^\top} \log f(y|\mathbf{x}, \boldsymbol{\eta})\| \leq M_4(y, \mathbf{x})$  for all  $\boldsymbol{\eta}$  satisfying  $\|\boldsymbol{\eta} - \boldsymbol{\eta}_0\| \leq \delta$ .

Define

$$\begin{aligned} \ell_{0*}(\boldsymbol{\beta}) &= \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0) \log \pi(\mathbf{X}^\top \boldsymbol{\beta}) + \{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\} \log \{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta})\}] \\ &= \mathbb{E}\{(\mathbf{X}^\top \boldsymbol{\beta}) \pi(\mathbf{X}^\top \boldsymbol{\beta}_0) - \log \{1 + \exp(\mathbf{X}^\top \boldsymbol{\beta})\}\}. \end{aligned} \quad (1)$$

This function is well defined for any  $\boldsymbol{\beta}$  under Condition (C1), because

$$|\ell_{0*}(\boldsymbol{\beta})| \leq \mathbb{E}\{|\mathbf{X}^\top \boldsymbol{\beta}| + \log \{1 + \exp(|\mathbf{X}^\top \boldsymbol{\beta}|)\}\} \leq 3\mathbb{E}\{|\mathbf{X}^\top \boldsymbol{\beta}|\} \leq 3\mathbb{E}(\|\mathbf{X}\|)\|\boldsymbol{\beta}\|,$$

and Condition (C1) implies  $\mathbb{E}\|\mathbf{X}\| < \infty$ .

**Theorem 1** Assume Conditions (C1)–(C5) and that  $H_0 : \gamma_0 = \gamma_1 = \theta_0 = \theta_1 = 0$  is true. As  $n$  goes to infinity,  $\sqrt{n}S_1(\widehat{\boldsymbol{\beta}}, \widehat{\boldsymbol{\zeta}}, \widehat{\boldsymbol{\eta}}) \xrightarrow{d} \mathcal{N}(0, \Sigma_1)$ , where

$$\Sigma_1 = \begin{pmatrix} \sigma_{11}^2 & \sigma_{12}^2 & \sigma_{13}^2 & \sigma_{14}^2 \\ \sigma_{21}^2 & \sigma_{22}^2 & \sigma_{23}^2 & \sigma_{24}^2 \\ \sigma_{31}^2 & \sigma_{32}^2 & \sigma_{33}^2 & \sigma_{34}^2 \\ \sigma_{41}^2 & \sigma_{42}^2 & \sigma_{43}^2 & \sigma_{44}^2 \end{pmatrix},$$

and

$$\begin{aligned}
\sigma_{11}^2 &= A_{21} + B_{21} - \mathbf{A}_{11}^\top \mathbf{A}^{-1} \mathbf{A}_{11} - \mathbf{B}_{11}^\top \mathbf{C}^{-1} \mathbf{B}_{11}, \\
\sigma_{22}^2 &= A_{22} + B_{22} - \mathbf{A}_{12}^\top \mathbf{A}^{-1} \mathbf{A}_{12} - \mathbf{B}_{12}^\top \mathbf{B}^{-1} \mathbf{B}_{12}, \\
\sigma_{33}^2 &= A_{23} + B_{23} - \mathbf{A}_{13}^\top \mathbf{A}^{-1} \mathbf{A}_{13} - \mathbf{B}_{13}^\top \mathbf{C}^{-1} \mathbf{B}_{13}, \\
\sigma_{44}^2 &= A_{24} + B_{24} - \mathbf{A}_{14}^\top \mathbf{A}^{-1} \mathbf{A}_{14} - \mathbf{B}_{14}^\top \mathbf{B}^{-1} \mathbf{B}_{14}, \\
\sigma_{12}^2 &= \sigma_{21}^2 = A_2 - \mathbf{A}_{11}^\top \mathbf{A}^{-1} \mathbf{A}_{12}, \\
\sigma_{13}^2 &= \sigma_{31}^2 = A_{31} + \mathbf{B}_{31} - \mathbf{A}_{11}^\top \mathbf{A}^{-1} \mathbf{A}_{13} - \mathbf{B}_{11}^\top \mathbf{C}^{-1} \mathbf{B}_{13}, \\
\sigma_{14}^2 &= \sigma_{41}^2 = A_3 - \mathbf{A}_{11}^\top \mathbf{A}^{-1} \mathbf{A}_{14}, \\
\sigma_{23}^2 &= \sigma_{32}^2 = A_4 - \mathbf{A}_{12}^\top \mathbf{A}^{-1} \mathbf{A}_{13}, \\
\sigma_{24}^2 &= \sigma_{42}^2 = A_{32} + \mathbf{B}_{32} - \mathbf{A}_{12}^\top \mathbf{A}^{-1} \mathbf{A}_{14} - \mathbf{B}_{12}^\top \mathbf{B}^{-1} \mathbf{B}_{14}, \\
\sigma_{34}^2 &= \sigma_{43}^2 = A_5 - \mathbf{A}_{13}^\top \mathbf{A}^{-1} \mathbf{A}_{14},
\end{aligned}$$

and

$$\begin{aligned}
\mathbf{A}_{11} &= \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}Y(0)\mathbf{X}], \\
\mathbf{A}_{12} &= \mathbb{E}\{\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}Y(1)\mathbf{X}\}, \\
\mathbf{A}_{13} &= \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}Y^2(0)\mathbf{X}], \\
\mathbf{A}_{14} &= \mathbb{E}\{\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}Y^2(1)\mathbf{X}\}, \\
A_{21} &= \mathbb{E}[\pi^2(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}Y^2(0)], \\
A_{22} &= \mathbb{E}\{\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2Y^2(1)\}, \\
A_{23} &= \mathbb{E}[\pi^2(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2Y^4(0)], \\
A_{24} &= \mathbb{E}\{\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2Y^4(1)\}, \\
A_{31} &= \mathbb{E}[\pi^2(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}Y^3(0)], \\
A_{32} &= \mathbb{E}\{\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}Y^3(1)\}, \\
A_2 &= \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}Y(0)Y(1)], \\
A_3 &= \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}Y(0)Y^2(1)], \\
A_4 &= \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}Y^2(0)Y(1)], \\
A_5 &= \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2Y^2(0)Y^2(1)], \\
\mathbf{B}_{11} &= \mathbb{E}\left[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\} \int y \nabla_{\boldsymbol{\eta}} f\{y|\mathbf{X}, \boldsymbol{\eta}_0\} dy\right], \\
\mathbf{B}_{12} &= \mathbb{E}\left[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\} \int y \nabla_{\boldsymbol{\zeta}} f\{y|\mathbf{X}, \boldsymbol{\zeta}_0\} dy\right], \\
\mathbf{B}_{13} &= \mathbb{E}\left[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\} \int y^2 \nabla_{\boldsymbol{\eta}} f\{y|\mathbf{X}, \boldsymbol{\eta}_0\} dy\right], \\
\mathbf{B}_{14} &= \mathbb{E}\left[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\} \int y^2 \nabla_{\boldsymbol{\zeta}} f\{y|\mathbf{X}, \boldsymbol{\zeta}_0\} dy\right], \\
B_{21} &= \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2 \mathbb{E}^2(Y(0)|X, \boldsymbol{\eta}_0)], \\
B_{22} &= \mathbb{E}[\{\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\} \mathbb{E}^2(Y(1)|X, \boldsymbol{\zeta}_0)], \\
B_{23} &= \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2 \mathbb{E}^2\{Y^2(0)|X, \boldsymbol{\eta}_0\}], \\
B_{24} &= \mathbb{E}[\{\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2 \mathbb{E}^2\{Y^2(1)|X, \boldsymbol{\zeta}_0\}], \\
B_{31} &= \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\} \mathbb{E}\{Y(0)|X, \boldsymbol{\eta}_0\} \mathbb{E}\{Y^2(0)|X, \boldsymbol{\zeta}_0\}], \\
B_{32} &= \mathbb{E}[\{\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\} \mathbb{E}\{Y(1)|X, \boldsymbol{\zeta}_0\} \mathbb{E}\{Y^2(1)|X, \boldsymbol{\zeta}_0\}].
\end{aligned}$$

## 2 A lemma

The following lemma eases much of the burden in our proof of Theorem 1.

**Lemma 1** Assume Conditions (C1)–(C5) and that  $H_0 : \gamma_0 = \gamma_1 = \theta_0 = \theta_1 = 0$  is true. As  $n$  goes to infinity,

$$\sqrt{n}(\hat{\beta} - \beta_0) = \mathbf{A}^{-1}n^{-1/2} \sum_{i=1}^n \{w_i - \pi(\mathbf{x}_i^\top \beta_0)\} \mathbf{x}_i + o_p(1), \quad (2)$$

$$\sqrt{n}(\hat{\zeta} - \zeta_0) = \mathbf{B}^{-1}n^{-1/2} \sum_{i=1}^n w_i \nabla_{\zeta} \log f\{y_i(1)|\mathbf{x}_i, \zeta_0\} + o_p(1), \quad (3)$$

$$\sqrt{n}(\hat{\eta} - \eta_0) = \mathbf{C}^{-1}n^{-1/2} \sum_{i=1}^n (1 - w_i) \nabla_{\eta} \log f\{y_i(0)|\mathbf{x}_i, \eta_0\} + o_p(1), \quad (4)$$

and

$$\sqrt{n}\{(\hat{\beta} - \beta_0)^\top, (\hat{\zeta} - \zeta_0)^\top, (\hat{\eta} - \eta_0)^\top\}^\top \xrightarrow{d} \mathcal{N}(\mathbf{0}, \text{diag}(\mathbf{A}^{-1}, \mathbf{B}^{-1}, \mathbf{C}^{-1})),$$

where the matrices  $\mathbf{A}$ ,  $\mathbf{B}$  and  $\mathbf{C}$  are defined in Conditions (C1), (C6) and (C7).

**Proof** By definition,  $\hat{\beta} = \arg \max_{\beta} \frac{1}{n} \ell_0(\beta)$ ,  $\hat{\zeta} = \arg \max_{\zeta} \frac{1}{n} \ell_1(\zeta)$ , and  $\hat{\eta} = \arg \max_{\eta} \frac{1}{n} \ell_2(\eta)$ , where  $\ell_0(\beta) = \sum_{i=1}^n [w_i \log \pi(\mathbf{x}_i^\top \beta) + (1 - w_i) \log \{1 - \pi(\mathbf{x}_i^\top \beta)\}]$ ,  $\ell_1(\zeta) = \sum_{i=1}^n \{w_i \log f\{y_i(1)|\mathbf{x}_i, \zeta\}\}$  and  $\ell_2(\eta) = \sum_{i=1}^n \{(1 - w_i) \log f\{y_i(0)|\mathbf{x}_i, \eta\}\}$ . Because  $\pi(\mathbf{x}_i^\top \beta)$ ,  $f\{y_i(1)|\mathbf{x}_i, \zeta\}$  and  $f\{y_i(0)|\mathbf{x}_i, \eta\}$  are differentiable with respect to the underlying parameters, equivalently the estimators  $\hat{\beta}$ ,  $\hat{\zeta}$  and  $\hat{\eta}$  satisfy

$$\sum_{i=1}^n \{w_i - \pi(\mathbf{x}_i^\top \hat{\beta})\} \mathbf{x}_i = 0, \quad (5)$$

$$\sum_{i=1}^n w_i \nabla_{\zeta} \log f\{y_i(1)|\mathbf{x}_i, \hat{\zeta}\} = 0, \quad (6)$$

$$\sum_{i=1}^n (1 - w_i) \nabla_{\eta} \log f\{y_i(0)|\mathbf{x}_i, \hat{\eta}\} = 0. \quad (7)$$

Both  $\ell_0(\beta)$  and  $\ell_{0*}(\beta)$  are concave functions, and  $(1/n)\ell_0(\beta)$  converges to  $\ell_{0*}(\beta)$  almost surely for each fixed  $\beta$ . By the convexity lemma of ?,  $\sup_{\beta} |(1/n)\ell_0(\beta) - \ell_{0*}(\beta)| = o_p(1)$ . Condition (C1) implies that  $\ell_{0*}(\beta)$  is strictly concave at  $\beta_0$ . This, together with  $\nabla_{\beta} \ell_{0*}(\beta)|_{\beta=\beta_0} = 0$ , implies that  $\beta_0$  is the unique maximizer of  $\ell_{0*}(\beta)$ . Then by Theorem 5.7 of ?,  $\hat{\beta} - \beta_0 = o_p(1)$ . The linear approximation (2) follows by first-order Taylor expansion of the left-hand side of (5) at  $\beta_0$  and by Condition (C1).

Next, we derive a linear approximation of  $\hat{\zeta}$  and  $\hat{\eta}$ . Let  $\ell_{1*}(\zeta) = \mathbb{E}[\pi(\mathbf{X}^\top \beta_0) \log f\{Y(1)|\mathbf{X}, \zeta_0\}]$  and  $\ell_{2*}(\eta) = \mathbb{E}[\{1 - \pi(\mathbf{X}^\top \beta_0)\} \log f\{Y(0)|\mathbf{X}, \eta_0\}]$ . The identifiability in Condition (C4) guarantees that  $\zeta_0$  is the unique maximizer of  $\ell_{1*}(\zeta)$ . Under Condition (C4),  $f\{y(1)|\mathbf{x}, \zeta\}$  is continuous in  $\zeta$  for almost all  $(y, \mathbf{x})$ . By Lemma 3.10 of ?, the class  $\{\log f\{y(1)|\mathbf{x}, \zeta\} : \zeta \in \Omega_1\}$  satisfies the uniform

law of large numbers, in other words,  $\sup_{\zeta \in \Omega_1} |\ell_1(\zeta)/n - \ell_{1*}(\zeta)| = o_p(1)$ . Therefore, the conditions of Theorem 5.7 of ? are met, and  $\hat{\zeta} - \zeta_0 = o_p(1)$ . The linear approximation (3) can be derived using similar techniques to those deriving the approximation (2). The linear approximation of  $\hat{\eta}$  can be derived using similar processes to those deriving the approximation of  $\hat{\zeta}$ .

The asymptotic result follows immediately from the linear approximations in (2), (3) and (4) by the central limit theorem. Condition (C6) and (C7) guarantees that the remainder terms in the linear approximations are negligible and that their asymptotic variances are well defined. This completes the proof of Lemma 1.

### 3 Proof of Theorem 1

We have shown that  $(\hat{\beta}, \hat{\zeta}, \hat{\eta})$  is  $\sqrt{n}$  consistent with  $(\beta_0, \zeta_0, \eta_0)$ . By first-order Taylor expansion of  $S_1(\hat{\beta}, \hat{\zeta}, \hat{\eta})$  at  $(\beta_0, \zeta_0, \eta_0)$ , we have

$$\begin{aligned} \sqrt{n}S_1(\hat{\beta}, \hat{\zeta}, \hat{\eta}) &= \sqrt{n}S_1(\beta_0, \zeta_0, \eta_0) + \{\nabla_{\beta}S_1(\tilde{\beta}, \tilde{\zeta}, \tilde{\eta})\}^{\top} \cdot \sqrt{n}(\hat{\beta} - \beta_0) \\ &\quad + \{\nabla_{\zeta}S_1(\tilde{\beta}, \tilde{\zeta}, \tilde{\eta})\}^{\top} \cdot \sqrt{n}(\hat{\zeta} - \zeta_0) \\ &\quad + \{\nabla_{\eta}S_1(\tilde{\beta}, \tilde{\zeta}, \tilde{\eta})\}^{\top} \cdot \sqrt{n}(\hat{\eta} - \eta_0), \end{aligned} \quad (8)$$

where  $(\tilde{\beta}, \tilde{\zeta}, \tilde{\eta})$  lies between  $(\beta_0, \zeta_0, \eta_0)$  and  $(\hat{\beta}, \hat{\zeta}, \hat{\eta})$ .

To proceed, we first need to derive the leading terms of  $\nabla_{\beta}S_1(\tilde{\beta}, \tilde{\zeta}, \tilde{\eta})$ ,  $\nabla_{\zeta}S_1(\tilde{\beta}, \tilde{\zeta}, \tilde{\eta})$  and  $\nabla_{\eta}S_1(\tilde{\beta}, \tilde{\zeta}, \tilde{\eta})$ . It follows from

$$S_1(\beta, \zeta, \eta) = \frac{1}{n} \sum_{i=1}^n \begin{pmatrix} w_i\{1 - \pi(\mathbf{x}_i^{\top}\beta)\}\mathbb{E}\{Y(0)|\mathbf{x}_i, \eta\} - (1 - w_i)\pi(\mathbf{x}_i^{\top}\beta)y_i(0) \\ w_i\{1 - \pi(\mathbf{x}_i^{\top}\beta)\}y_i(1) - (1 - w_i)\pi(\mathbf{x}_i^{\top}\beta)\mathbb{E}\{Y(1)|\mathbf{x}_i, \zeta\} \\ w_i\{1 - \pi(\mathbf{x}_i^{\top}\beta)\}\{1 - 2\pi(\mathbf{x}_i^{\top}\beta)\}\mathbb{E}\{Y^2(0)|\mathbf{x}_i, \eta\} - (1 - w_i)\pi(\mathbf{x}_i^{\top}\beta)\{1 - 2\pi(\mathbf{x}_i^{\top}\beta)\}y_i^2(0) \\ w_i\{1 - \pi(\mathbf{x}_i^{\top}\beta)\}\{1 - 2\pi(\mathbf{x}_i^{\top}\beta)\}y_i^2(1) - (1 - w_i)\pi(\mathbf{x}_i^{\top}\beta)\{1 - 2\pi(\mathbf{x}_i^{\top}\beta)\}\mathbb{E}\{Y^2(1)|\mathbf{x}_i, \zeta\} \end{pmatrix},$$

that

$$\begin{aligned} \nabla_{\beta^{\top}}S_1(\beta, \zeta, \eta) &= \frac{1}{n} \sum_{i=1}^n \begin{pmatrix} -w_i\pi(\mathbf{x}_i^{\top}\beta)\{1 - \pi(\mathbf{x}_i^{\top}\beta)\}\mathbb{E}\{Y(0)|\mathbf{x}_i, \eta\}\mathbf{x}_i^{\top} \\ -w_i\pi(\mathbf{x}_i^{\top}\beta)\{1 - \pi(\mathbf{x}_i^{\top}\beta)\}y_i(1)\mathbf{x}_i^{\top} \\ -w_i\pi(\mathbf{x}_i^{\top}\beta)\{1 - \pi(\mathbf{x}_i^{\top}\beta)\}\{3 - 4\pi(\mathbf{x}_i^{\top}\beta)\}\mathbb{E}\{Y^2(0)|\mathbf{x}_i, \eta\}\mathbf{x}_i^{\top} \\ -w_i\pi(\mathbf{x}_i^{\top}\beta)\{1 - \pi(\mathbf{x}_i^{\top}\beta)\}\{3 - 4\pi(\mathbf{x}_i^{\top}\beta)\}y_i^2(1)\mathbf{x}_i^{\top} \end{pmatrix} \\ &\quad - \frac{1}{n} \sum_{i=1}^n \begin{pmatrix} (1 - w_i)\pi(\mathbf{x}_i^{\top}\beta)\{1 - \pi(\mathbf{x}_i^{\top}\beta)\}y_i(0)\mathbf{x}_i^{\top} \\ (1 - w_i)\pi(\mathbf{x}_i^{\top}\beta)\{1 - \pi(\mathbf{x}_i^{\top}\beta)\}\mathbb{E}\{Y(1)|\mathbf{x}_i, \zeta\}\mathbf{x}_i^{\top} \\ (1 - w_i)\pi(\mathbf{x}_i^{\top}\beta)\{1 - \pi(\mathbf{x}_i^{\top}\beta)\}\{1 - 4\pi(\mathbf{x}_i^{\top}\beta)\}y_i^2(0)\mathbf{x}_i^{\top} \\ (1 - w_i)\pi(\mathbf{x}_i^{\top}\beta)\{1 - \pi(\mathbf{x}_i^{\top}\beta)\}\{1 - 4\pi(\mathbf{x}_i^{\top}\beta)\}\mathbb{E}\{Y^2(1)|\mathbf{x}_i, \zeta\}\mathbf{x}_i^{\top} \end{pmatrix}, \end{aligned}$$

$$\begin{aligned}\nabla_{\zeta^\top} S_1(\beta, \zeta, \eta) &= \frac{1}{n} \sum_{i=1}^n \begin{pmatrix} \mathbf{0}^\top \\ -(1-w_i)\pi(\mathbf{x}_i^\top \beta) \nabla_{\zeta^\top} \mathbb{E}\{Y(1)|\mathbf{x}_i, \zeta\} \\ \mathbf{0}^\top \\ -(1-w_i)\pi(\mathbf{x}_i^\top \beta) \{1 - 2\pi(\mathbf{x}_i^\top \beta)\} \nabla_{\zeta^\top} \mathbb{E}\{Y^2(1)|\mathbf{x}_i, \zeta\} \end{pmatrix}, \\ \nabla_{\eta^\top} S_1(\beta, \zeta, \eta) &= \frac{1}{n} \sum_{i=1}^n \begin{pmatrix} w_i \{1 - \pi(\mathbf{x}_i^\top \beta)\} \nabla_{\eta^\top} \mathbb{E}\{Y(0)|\mathbf{x}_i, \eta\} \\ \mathbf{0}^\top \\ w_i \{1 - \pi(\mathbf{x}_i^\top \beta)\} \{1 - 2\pi(\mathbf{x}_i^\top \beta)\} \nabla_{\eta^\top} \mathbb{E}\{Y^2(0)|\mathbf{x}_i, \eta\} \\ \mathbf{0}^\top \end{pmatrix}.\end{aligned}$$

Under Conditions (C1), (C6) and (C7), by the weak law of large numbers, they converge in probability respectively to

$$\begin{aligned}\mathbb{E}[\nabla_{\beta^\top} S_1(\beta, \zeta, \eta)] &= \mathbb{E} \begin{pmatrix} -\pi(\mathbf{X}^\top \beta_0) \pi(\mathbf{X}^\top \beta) \{1 - \pi(\mathbf{X}^\top \beta)\} \mathbb{E}\{Y(0)|\mathbf{X}, \eta\} \mathbf{X}^\top \\ -\pi(\mathbf{X}^\top \beta_0) \pi(\mathbf{X}^\top \beta) \{1 - \pi(\mathbf{X}^\top \beta)\} Y(1) \mathbf{X}^\top \\ -\pi(\mathbf{X}^\top \beta_0) \pi(\mathbf{X}^\top \beta) \{1 - \pi(\mathbf{X}^\top \beta)\} \{3 - 4\pi(\mathbf{X}^\top \beta)\} \mathbb{E}\{Y^2(0)|\mathbf{X}, \eta\} \mathbf{X}^\top \\ -\pi(\mathbf{X}^\top \beta_0) \pi(\mathbf{X}^\top \beta) \{1 - \pi(\mathbf{X}^\top \beta)\} \{3 - 4\pi(\mathbf{X}^\top \beta)\} Y^2(1) \mathbf{X}^\top \end{pmatrix} \\ &\quad - \mathbb{E} \begin{pmatrix} \{1 - \pi(\mathbf{X}^\top \beta_0)\} \pi(\mathbf{X}^\top \beta) \{1 - \pi(\mathbf{X}^\top \beta)\} Y(0) \mathbf{X}^\top \\ \{1 - \pi(\mathbf{X}^\top \beta_0)\} \pi(\mathbf{X}^\top \beta) \{1 - \pi(\mathbf{X}^\top \beta)\} \mathbb{E}\{Y(1)|\mathbf{X}, \zeta\} \mathbf{X}^\top \\ \{1 - \pi(\mathbf{X}^\top \beta_0)\} \pi(\mathbf{X}^\top \beta) \{1 - \pi(\mathbf{X}^\top \beta)\} \{1 - 4\pi(\mathbf{X}^\top \beta)\} Y^2(0) \mathbf{X}^\top \\ \{1 - \pi(\mathbf{X}^\top \beta_0)\} \pi(\mathbf{X}^\top \beta) \{1 - \pi(\mathbf{X}^\top \beta)\} \{1 - 4\pi(\mathbf{X}^\top \beta)\} \mathbb{E}\{Y^2(1)|\mathbf{X}, \zeta\} \mathbf{X}^\top \end{pmatrix}, \\ \mathbb{E}[\nabla_{\zeta^\top} S_1(\beta, \zeta, \eta)] &= \mathbb{E} \begin{pmatrix} \mathbf{0}^\top \\ -\{1 - \pi(\mathbf{X}^\top \beta_0)\} \pi(\mathbf{X}^\top \beta) \nabla_{\zeta^\top} \mathbb{E}\{Y(1)|\mathbf{X}, \zeta\} \\ \mathbf{0}^\top \\ -\{1 - \pi(\mathbf{X}^\top \beta_0)\} \pi(\mathbf{X}^\top \beta) \{1 - 2\pi(\mathbf{X}^\top \beta)\} \nabla_{\zeta^\top} \mathbb{E}\{Y^2(1)|\mathbf{X}, \zeta\} \end{pmatrix}, \\ \mathbb{E}[\nabla_{\eta^\top} S_1(\beta, \zeta, \eta)] &= \mathbb{E} \begin{pmatrix} \pi(\mathbf{X}^\top \beta_0) \{1 - \pi(\mathbf{X}^\top \beta)\} \nabla_{\eta^\top} \mathbb{E}\{Y(0)|\mathbf{X}, \eta\} \\ \mathbf{0}^\top \\ \pi(\mathbf{X}^\top \beta_0) \{1 - \pi(\mathbf{X}^\top \beta)\} \{1 - 2\pi(\mathbf{X}^\top \beta)\} \nabla_{\eta^\top} \mathbb{E}\{Y^2(0)|\mathbf{X}, \eta\} \\ \mathbf{0}^\top \end{pmatrix}.\end{aligned}$$

Conditions (C1), (C6) and (C7) also imply that the above convergences hold uniformly in  $(\beta, \zeta, \eta)$ , and therefore the two limit functions are continuous functions of  $(\beta, \zeta, \eta)$ . This, together with the

consistency of  $(\widehat{\boldsymbol{\beta}}, \widehat{\boldsymbol{\zeta}}, \widehat{\boldsymbol{\eta}})$ , leads to

$$\begin{aligned}\nabla_{\boldsymbol{\beta}^\top} S_1(\widetilde{\boldsymbol{\beta}}, \widetilde{\boldsymbol{\zeta}}, \widetilde{\boldsymbol{\eta}}) &= - \begin{pmatrix} \mathbf{A}_{11}^\top \\ \mathbf{A}_{12}^\top \\ \mathbf{A}_{13}^\top \\ \mathbf{A}_{14}^\top \end{pmatrix} + o_p(1), \\ \nabla_{\boldsymbol{\zeta}^\top} S_1(\widetilde{\boldsymbol{\beta}}, \widetilde{\boldsymbol{\zeta}}, \widetilde{\boldsymbol{\eta}}) &= - \begin{pmatrix} \mathbf{0}^\top \\ \mathbf{B}_{12}^\top \\ \mathbf{0}^\top \\ \mathbf{B}_{14}^\top \end{pmatrix} + o_p(1), \\ \nabla_{\boldsymbol{\eta}^\top} S_1(\widetilde{\boldsymbol{\beta}}, \widetilde{\boldsymbol{\zeta}}, \widetilde{\boldsymbol{\eta}}) &= \begin{pmatrix} \mathbf{B}_{11}^\top \\ \mathbf{0}^\top \\ \mathbf{B}_{13}^\top \\ \mathbf{0}^\top \end{pmatrix} + o_p(1),\end{aligned}$$

where

$$\begin{aligned}\mathbf{A}_{11} &= \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}Y(0)\mathbf{X}], \\ \mathbf{A}_{12} &= \mathbb{E}\{\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}Y(1)\mathbf{X}\}, \\ \mathbf{A}_{13} &= \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}Y^2(0)\mathbf{X}], \\ \mathbf{A}_{14} &= \mathbb{E}\{\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}Y^2(1)\mathbf{X}\}, \\ \mathbf{B}_{11} &= \mathbb{E}\left[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\} \int y \nabla_{\boldsymbol{\eta}} f\{y|\mathbf{X}, \boldsymbol{\eta}_0\} dy\right], \\ \mathbf{B}_{12} &= \mathbb{E}\left[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\} \int y \nabla_{\boldsymbol{\zeta}} f\{y|\mathbf{X}, \boldsymbol{\zeta}_0\} dy\right], \\ \mathbf{B}_{13} &= \mathbb{E}\left[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\} \int y^2 \nabla_{\boldsymbol{\eta}} f\{y|\mathbf{X}, \boldsymbol{\eta}_0\} dy\right], \\ \mathbf{B}_{14} &= \mathbb{E}\left[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}\{1 - 2\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\} \int y^2 \nabla_{\boldsymbol{\zeta}} f\{y|\mathbf{X}, \boldsymbol{\zeta}_0\} dy\right].\end{aligned}$$

Consequently, the equality (8) reduces to

$$\begin{aligned}\sqrt{n}S_1(\widehat{\boldsymbol{\beta}}, \widehat{\boldsymbol{\zeta}}, \widehat{\boldsymbol{\eta}}) &= \sqrt{n}S_1(\boldsymbol{\beta}_0, \boldsymbol{\zeta}_0, \boldsymbol{\eta}_0) - \begin{pmatrix} \mathbf{A}_{11}^\top \\ \mathbf{A}_{12}^\top \\ \mathbf{A}_{13}^\top \\ \mathbf{A}_{14}^\top \end{pmatrix} \sqrt{n}(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}_0) - \begin{pmatrix} \mathbf{0}^\top \\ \mathbf{B}_{12}^\top \\ \mathbf{0}^\top \\ \mathbf{B}_{14}^\top \end{pmatrix} \sqrt{n}(\widehat{\boldsymbol{\zeta}} - \boldsymbol{\zeta}_0) \\ &\quad + \begin{pmatrix} \mathbf{B}_{11}^\top \\ \mathbf{0}^\top \\ \mathbf{B}_{13}^\top \\ \mathbf{0}^\top \end{pmatrix} \sqrt{n}(\widehat{\boldsymbol{\eta}} - \boldsymbol{\eta}_0) + o_p(1).\end{aligned}\tag{9}$$



Putting the linear approximations of  $\widehat{\beta}$ ,  $\widehat{\zeta}$  and  $\widehat{\eta}$  from (2), (3) and (4) into (9), we have

$$\sqrt{n}S_1(\widehat{\beta}, \widehat{\zeta}, \widehat{\eta}) = \frac{1}{\sqrt{n}} \sum_{i=1}^n \begin{pmatrix} v_{i1} + v_{i2} + v_{i3} + v_{i4} \\ v_{i5} + v_{i6} + v_{i7} + v_{i8} \\ v_{i9} + v_{i10} + v_{i11} + v_{i12} \\ v_{i13} + v_{i14} + v_{i15} + v_{i16} \end{pmatrix} + o_p(1),$$

where

$$\begin{aligned} v_{i1} &= w_i \{1 - \pi(\mathbf{x}_i^\top \beta_0)\} \mathbb{E}\{Y(0)|\mathbf{x}_i, \boldsymbol{\eta}_0\}, & v_{i2} &= -(1 - w_i) \pi(\mathbf{x}_i^\top \beta_0) y_i(0), \\ v_{i3} &= -\mathbf{A}_{11}^\top \mathbf{A}^{-1} \{w_i - \pi(\mathbf{x}_i^\top \beta_0)\} \mathbf{x}_i, & v_{i4} &= \mathbf{B}_{11}^\top \mathbf{C}^{-1} (1 - w_i) \nabla_{\boldsymbol{\eta}} \log f\{y_i(0)|\mathbf{x}_i, \boldsymbol{\eta}_0\}, \\ v_{i5} &= w_i \{1 - \pi(\mathbf{x}_i^\top \beta_0)\} y_i(1), & v_{i6} &= -(1 - w_i) \pi(\mathbf{x}_i^\top \beta_0) \mathbb{E}\{Y(1)|\mathbf{x}_i, \zeta_0\}, \\ v_{i7} &= -\mathbf{A}_{12}^\top \mathbf{A}^{-1} \{w_i - \pi(\mathbf{x}_i^\top \beta_0)\} \mathbf{x}_i, & v_{i8} &= -\mathbf{B}_{12}^\top \mathbf{B}^{-1} w_i \nabla_{\zeta} \log f(y_i(1)|\mathbf{x}_i, \zeta_0), \\ v_{i9} &= w_i \{1 - \pi(\mathbf{x}_i^\top \beta_0)\} \{1 - 2\pi(\mathbf{x}_i^\top \beta_0)\} \mathbb{E}\{Y^2(0)|\mathbf{x}_i, \boldsymbol{\eta}_0\}, & v_{i10} &= -(1 - w_i) \pi(\mathbf{x}_i^\top \beta_0) \{1 - 2\pi(\mathbf{x}_i^\top \beta_0)\} y_i^2(0), \\ v_{i11} &= -\mathbf{A}_{13}^\top \mathbf{A}^{-1} \{w_i - \pi(\mathbf{x}_i^\top \beta_0)\} \mathbf{x}_i, & v_{i12} &= \mathbf{B}_{13}^\top \mathbf{C}^{-1} (1 - w_i) \nabla_{\boldsymbol{\eta}} \log f\{y_i(0)|\mathbf{x}_i, \boldsymbol{\eta}_0\}, \\ v_{i13} &= w_i \{1 - \pi(\mathbf{x}_i^\top \beta_0)\} \{1 - 2\pi(\mathbf{x}_i^\top \beta_0)\} y_i^2(1), & v_{i14} &= -(1 - w_i) \pi(\mathbf{x}_i^\top \beta_0) \{1 - 2\pi(\mathbf{x}_i^\top \beta_0)\} \mathbb{E}\{Y^2(1)|\mathbf{x}_i, \zeta_0\}, \\ v_{i15} &= -\mathbf{A}_{14}^\top \mathbf{A}^{-1} \{w_i - \pi(\mathbf{x}_i^\top \beta_0)\} \mathbf{x}_i, & v_{i16} &= -\mathbf{B}_{14}^\top \mathbf{B}^{-1} w_i \nabla_{\zeta} \log f(y_i(1)|\mathbf{x}_i, \zeta_0). \end{aligned}$$

By the central limit theorem, as  $n \rightarrow \infty$ ,  $\sqrt{n}S_1(\widehat{\beta}, \widehat{\zeta}, \widehat{\eta}) \xrightarrow{d} N(0, \Sigma_1)$ , with

$$\Sigma_1 = \text{Var} \begin{pmatrix} v_{i1} + v_{i2} + v_{i3} + v_{i4} \\ v_{i5} + v_{i6} + v_{i7} + v_{i8} \\ v_{i9} + v_{i10} + v_{i11} + v_{i12} \\ v_{i13} + v_{i14} + v_{i15} + v_{i16} \end{pmatrix}.$$

Because  $\mathbb{E}(v_{i1} + v_{i2}) = \mathbb{E}(v_{i5} + v_{i6}) = \mathbb{E}(v_{i9} + v_{i10}) = \mathbb{E}(v_{i13} + v_{i14}) = \mathbb{E}(v_{i3}) = \mathbb{E}(v_{i4}) = \mathbb{E}(v_{i7}) = \mathbb{E}(v_{i8}) = \mathbb{E}(v_{i11}) = \mathbb{E}(v_{i12}) = \mathbb{E}(v_{i15}) = \mathbb{E}(v_{i16}) = 0$ , we have

$$\begin{aligned} \sigma_{11}^2 &= \mathbb{E}(v_{i1} + v_{i2} + v_{i3} + v_{i4})^2, & \sigma_{22}^2 &= \mathbb{E}(v_{i5} + v_{i6} + v_{i7} + v_{i8})^2, \\ \sigma_{33}^2 &= \mathbb{E}(v_{i9} + v_{i10} + v_{i11} + v_{i12})^2, & \sigma_{44}^2 &= \mathbb{E}(v_{i13} + v_{i14} + v_{i15} + v_{i16})^2, \\ \sigma_{12}^2 &= \sigma_{21}^2 = \mathbb{E}\{(v_{i1} + v_{i2} + v_{i3} + v_{i4})(v_{i5} + v_{i6} + v_{i7} + v_{i8})\}, \\ \sigma_{13}^2 &= \sigma_{31}^2 = \mathbb{E}\{(v_{i1} + v_{i2} + v_{i3} + v_{i4})(v_{i9} + v_{i10} + v_{i11} + v_{i12})\}, \\ \sigma_{14}^2 &= \sigma_{41}^2 = \mathbb{E}\{(v_{i1} + v_{i2} + v_{i3} + v_{i4})(v_{i13} + v_{i14} + v_{i15} + v_{i16})\}, \\ \sigma_{23}^2 &= \sigma_{32}^2 = \mathbb{E}\{(v_{i5} + v_{i6} + v_{i7} + v_{i8})(v_{i9} + v_{i10} + v_{i11} + v_{i12})\}, \\ \sigma_{24}^2 &= \sigma_{42}^2 = \mathbb{E}\{(v_{i5} + v_{i6} + v_{i7} + v_{i8})(v_{i13} + v_{i14} + v_{i15} + v_{i16})\}, \\ \sigma_{34}^2 &= \sigma_{43}^2 = \mathbb{E}\{(v_{i9} + v_{i10} + v_{i11} + v_{i12})(v_{i13} + v_{i14} + v_{i15} + v_{i16})\}. \end{aligned}$$

With tedious algebra, it can be found that

$$\begin{aligned}\mathbb{E}(v_{i1}^2) &= \mathbb{E} \left[ \pi(\mathbf{X}^\top \boldsymbol{\beta}_0) \{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2 \left\{ \int y f_0(y|X, \eta_0) dy \right\}^2 \right] = B_{21}, \\ \mathbb{E}(v_{i2}^2) &= \mathbb{E} [\pi^2(\mathbf{X}^\top \boldsymbol{\beta}_0) \{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\} Y^2(0)] = A_{21}, \\ \mathbb{E}(v_{i3}^2) &= \mathbf{A}_{11}^\top \mathbf{A}^{-1} \mathbf{A}_{11}, \quad \mathbb{E}(v_{i4}^2) = \mathbf{B}_{11}^\top \mathbf{C}^{-1} \mathbf{B}_{11}\end{aligned}$$

and that  $\mathbb{E}(v_{i1}v_{i2}) = \mathbb{E}(v_{i1}v_{i4}) = \mathbb{E}(v_{i3}v_{i4}) = 0$ ,

$$\begin{aligned}\mathbb{E}(v_{i1}v_{i3}) &= -\mathbf{A}_{11}^\top \mathbf{A}^{-1} \mathbb{E}[\pi(\mathbf{X}^\top \boldsymbol{\beta}_0) \{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2 \mathbf{X} Y(0)], \\ \mathbb{E}(v_{i2}v_{i3}) &= -\mathbf{A}_{11}^\top \mathbf{A}^{-1} \mathbb{E} [\{\pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\}^2 \{1 - \pi(\mathbf{X}^\top \boldsymbol{\beta}_0)\} \mathbf{X} Y(0)], \\ \mathbb{E}(v_{i2}v_{i4}) &= -\mathbf{B}_{11}^\top \mathbf{C}^{-1} \mathbf{B}_{11}.\end{aligned}$$

Note that  $\mathbb{E}(v_{i1}v_{i3}) + \mathbb{E}(v_{i2}v_{i3}) = -\mathbf{A}_{11}^\top \mathbf{A}^{-1} \mathbf{A}_{11}$ . In summary, we have

$$\sigma_{11}^2 = A_{21} + B_{21} - \mathbf{A}_{11}^\top \mathbf{A}^{-1} \mathbf{A}_{11} - \mathbf{B}_{11}^\top \mathbf{C}^{-1} \mathbf{B}_{11}.$$

With the same tedious algebra process as  $\sigma_{11}^2$ , we have

$$\begin{aligned}\sigma_{22}^2 &= A_{22} + B_{22} - \mathbf{A}_{12}^\top \mathbf{A}^{-1} \mathbf{A}_{12} - \mathbf{B}_{12}^\top \mathbf{B}^{-1} \mathbf{B}_{12}, \\ \sigma_{33}^2 &= A_{23} + B_{23} - \mathbf{A}_{13}^\top \mathbf{A}^{-1} \mathbf{A}_{13} - \mathbf{B}_{13}^\top \mathbf{C}^{-1} \mathbf{B}_{13}, \\ \sigma_{44}^2 &= A_{24} + B_{24} - \mathbf{A}_{14}^\top \mathbf{A}^{-1} \mathbf{A}_{14} - \mathbf{B}_{14}^\top \mathbf{B}^{-1} \mathbf{B}_{14}, \\ \sigma_{12}^2 &= \sigma_{21}^2 = A_2 - \mathbf{A}_{11}^\top \mathbf{A}^{-1} \mathbf{A}_{12}, \\ \sigma_{13}^2 &= \sigma_{31}^2 = A_{31} + \mathbf{B}_{31} - \mathbf{A}_{11}^\top \mathbf{A}^{-1} \mathbf{A}_{13} - \mathbf{B}_{11}^\top \mathbf{C}^{-1} \mathbf{B}_{13}, \\ \sigma_{14}^2 &= \sigma_{41}^2 = A_3 - \mathbf{A}_{11}^\top \mathbf{A}^{-1} \mathbf{A}_{14}, \\ \sigma_{23}^2 &= \sigma_{32}^2 = A_4 - \mathbf{A}_{12}^\top \mathbf{A}^{-1} \mathbf{A}_{13}, \\ \sigma_{24}^2 &= \sigma_{42}^2 = A_{32} + \mathbf{B}_{32} - \mathbf{A}_{12}^\top \mathbf{A}^{-1} \mathbf{A}_{14} - \mathbf{B}_{12}^\top \mathbf{B}^{-1} \mathbf{B}_{14}, \\ \sigma_{34}^2 &= \sigma_{43}^2 = A_5 - \mathbf{A}_{13}^\top \mathbf{A}^{-1} \mathbf{A}_{14}.\end{aligned}$$

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