

Sandwich Variance Estimator of DR ATE Estimator

: Reupdate Version (9월 20일 Version)

: Estimating equation 잘못 작성한 부분 수정

$$\psi_T(\theta) = \begin{bmatrix} v_1 - \frac{\{E_T Y_T - (E_T - e_T) E[Y_T | E=1, X_T]\}}{e_T} \\ v_0 - \frac{\{(1-E_T) Y_T + (E_T - e_T) E[Y_T | E=0, X_T]\}}{1 - e_T} \\ X_T (Y_T - \alpha^T X_T) \\ \underbrace{E_T \cdot X_T - \frac{e^{X_T^T \beta}}{1 + e^{X_T^T \beta}} \cdot X_T}_{\text{PS model의 model matrix}} \end{bmatrix} \begin{matrix} \dots (\psi_{v_1}) \\ \dots (\psi_{v_0}) \\ \dots (\psi_\alpha) \\ \dots (\psi_\beta) \end{matrix}$$

Covariance term

Outcome regression model의 model matrix

PS model의 model matrix

$\theta = (v_1, v_0, \alpha, \beta)$

$$\frac{\partial}{\partial \theta} \psi_T(\hat{\theta}) = \begin{bmatrix} \frac{\partial}{\partial v_1} \psi_{v_1}, \frac{\partial}{\partial v_0} \psi_{v_1}, \frac{\partial}{\partial \alpha} \psi_{v_1}, \frac{\partial}{\partial \beta} \psi_{v_1} \\ \frac{\partial}{\partial v_1} \psi_{v_0}, \frac{\partial}{\partial v_0} \psi_{v_0}, \frac{\partial}{\partial \alpha} \psi_{v_0}, \frac{\partial}{\partial \beta} \psi_{v_0} \\ \frac{\partial}{\partial v_1} \psi_\alpha, \frac{\partial}{\partial v_0} \psi_\alpha, \frac{\partial}{\partial \alpha} \psi_\alpha, \frac{\partial}{\partial \beta} \psi_\alpha \\ \frac{\partial}{\partial v_1} \psi_\beta, \frac{\partial}{\partial v_0} \psi_\beta, \frac{\partial}{\partial \alpha} \psi_\beta, \frac{\partial}{\partial \beta} \psi_\beta \end{bmatrix} \begin{matrix} \dots (1) \\ \dots (2) \\ \dots (3) \\ \dots (4) \end{matrix}$$

$$\begin{aligned} 1) \psi_{v_1} &= v_1 - \frac{(1 + e^{\beta^T X_T}) \{E_T Y_T - (E_T - \frac{e^{\beta^T X_T}}{1 + e^{\beta^T X_T}}) \cdot (1, 1, X_T^T) \alpha\}}{e^{\beta^T X_T}} \\ &= v_1 - \frac{(1 + e^{\beta^T X_T}) (E_T Y_T) - E_T (1 + e^{\beta^T X_T}) \cdot (1, 1, X_T^T) \alpha + e^{\beta^T X_T} \cdot (1, 1, X_T^T) \alpha}{e^{\beta^T X_T}} \\ &= v_1 - \frac{(1 + e^{\beta^T X_T}) E_T Y_T}{e^{\beta^T X_T}} + \frac{E_T (1 + e^{\beta^T X_T})}{e^{\beta^T X_T}} \cdot (1, 1, X_T^T) \alpha - (1, 1, X_T^T) \alpha \end{aligned}$$

$$\therefore \frac{\partial}{\partial v_1} \psi_{v_1} = 1 \quad / \quad \frac{\partial}{\partial v_0} \psi_{v_1} = 0$$

$$\therefore \frac{\partial}{\partial \alpha} \psi_{v_1} = \frac{E_T(1+e^{\beta^T X_T})}{e^{\beta^T X_T}} \cdot (1, 1, X_T^T) - (1, 1, X_T^T) = \left(\frac{E_T(1+e^{\beta^T X_T})}{e^{\beta^T X_T}} - 1 \right) (1, 1, X_T^T)$$

Our situation ; (B.T. CT)

$$\therefore \psi_{v_1} \text{에서 } \beta \text{와 관련된 부분} \Rightarrow -\frac{E_T Y_T}{e^{X_T^T \beta}} + \frac{E_T(1, 1, X_T^T) \alpha}{e^{X_T^T \beta}} - \frac{E_T(Y_T - (1, 1, X_T^T) \alpha)}{e^{X_T^T \beta}} =$$

$$\frac{\partial}{\partial \beta} \psi_{v_1} = \frac{E_T Y_T \cdot e^{\beta^T X_T} \cdot X_T}{(e^{\beta^T X_T})^2} - \frac{E_T(1, 1, X_T^T) \alpha \cdot e^{\beta^T X_T} \cdot X_T}{(e^{\beta^T X_T})^2} + \frac{E_T(Y_T - (1, 1, X_T^T) \alpha) e^{\beta^T X_T} \cdot X_T}{(e^{\beta^T X_T})^2}$$

$$= \frac{E_T(Y_T - (1, 1, X_T^T) \alpha + Y_T - (1, 1, X_T^T) \alpha) \cdot X_T}{e^{\beta^T X_T}} = \frac{2 \cdot E_T(Y_T - (1, 1, X_T^T) \alpha) \cdot X_T}{e^{\beta^T X_T}}$$

PS model의 model matrix

$$= v_0 - (1 - E_T) Y_T - e^{X_T^T \beta} (1 - E_T) Y_T - E_T \cdot (1, 0, X_T^T) \alpha - e^{X_T^T \beta} \cdot E_T \cdot (1, 0, X_T^T) \alpha + e^{X_T^T \beta} \cdot (1, 0, X_T^T) \alpha$$

$$2) \psi_{v_0} = v_0 - (1 + e^{\beta^T X_T}) (1 - E_T) Y_T - (1 + e^{\beta^T X_T}) \cdot E_T \cdot (1, 0, X_T^T) \alpha + e^{\beta^T X_T} \cdot (1, 0, X_T^T) \alpha$$

$$\therefore \frac{\partial}{\partial v_1} \psi_{v_0} = 0, \quad \frac{\partial}{\partial v_0} \psi_{v_0} = 1, \quad \frac{\partial}{\partial \alpha} \psi_{v_0} = -(1 + e^{\beta^T X_T}) \cdot E_T \cdot (1, 0, X_T^T) + e^{\beta^T X_T} (1, 0, X_T^T)$$

$$= -(1 + e^{\beta^T X_T}) \cdot E_T + e^{\beta^T X_T} (1, 0, X_T^T)$$

$$\therefore \psi_{v_0} \text{ 중 } \beta \text{와 관련된 부분} \Rightarrow -e^{\beta^T X_T} (1 - E_T) Y_T - e^{\beta^T X_T} \cdot E_T (1, 0, X_T^T) \alpha + e^{\beta^T X_T} (1, 0, X_T^T) \alpha$$

$$\frac{\partial}{\partial \beta} \psi_{v_0} = -e^{\beta^T X_T} (1 - E_T) Y_T \cdot X_T - e^{\beta^T X_T} \cdot E_T \cdot (1, 0, X_T^T) \alpha \cdot X_T + e^{\beta^T X_T} (1, 0, X_T^T) \alpha \cdot X_T$$

$$= e^{\beta^T X_T} \cdot X_T (- (1 - E_T) Y_T - E_T (1, 0, X_T^T) \alpha + (1, 0, X_T^T) \alpha)$$

PS model의 model matrix

$$= e^{\beta^T X_T} (- (1 - E_T) Y_T + (1, 0, X_T^T) \alpha (1 - E_T)) X_T = -e^{X_T^T \beta} ((1 - E_T) (Y_T - (1, 0, X_T^T) \alpha)) \cdot X_T$$

$$3) \psi_{\alpha} = X_T^T (Y_T - X_T^T \alpha) \quad (X_T \text{는 Outcome model의 model matrix} \Rightarrow (1, E, X))$$

$$\therefore \frac{\partial}{\partial v_1} \psi_{\alpha} = 0, \quad \frac{\partial}{\partial v_0} \psi_{\alpha} = 0, \quad \frac{\partial}{\partial \alpha} \psi_{\alpha} = -X_T X_T^T, \quad \frac{\partial}{\partial \beta} \psi_{\alpha} = 0 \quad (\leftarrow \text{원인 개수는 0, 결과의 개수})$$

$$4) \psi_{\beta} = E_T \cdot X_T - \frac{e^{X_T^T \beta}}{1 + e^{X_T^T \beta}} \cdot X_T \quad (X_T \text{는 PS model의 model matrix} \Rightarrow (1, X))$$

$$\therefore \frac{\partial}{\partial v_1} \psi_{\beta} = \frac{\partial}{\partial v_0} \psi_{\beta} = 0, \quad \frac{\partial}{\partial \alpha} \psi_{\beta} = 0 \quad (\leftarrow \text{원인 개수는 0, 결과의 개수})$$

$$\frac{\partial}{\partial \beta} \psi_{\beta} = \frac{-e^{X_T^T \beta}}{1 + e^{X_T^T \beta}} \cdot X_T \cdot X_T^T$$