

# # Sandwich Variance of DR ATT Estimator #

- 11월 4일 Version (수정 Version)

Estimating equation)

$$\psi_{\tau}(\theta) = \begin{bmatrix} v_1 - \frac{N_1}{N} \times (Y_{\tau} \cdot E_{\tau}) & \text{--- } \psi_{v_1} \\ v_0 - \frac{N_1}{N} \times \frac{\{Y_{\tau} \cdot (1 - E_{\tau}) \cdot e_{\tau} + E[Y_{\tau} | E=0, X_{c,\tau}] \cdot (E_{\tau} - e_{\tau})\}}{1 - e_{\tau}} & \text{--- } \psi_{v_0} \\ X_{0,\tau} (y_{\tau} - X_{0,\tau}^T \alpha) & \text{--- } \psi_{\alpha} \\ \left(E_{\tau} - \frac{e^{x_{p,\tau}^T \beta}}{1 + e^{x_{p,\tau}^T \beta}}\right) \cdot X_{p,\tau}^T & \text{--- } \psi_{\beta} \end{bmatrix}$$

$\theta = (v_1, v_0, \alpha^T, \beta^T)^T$  /  $X_c$  = Covariates Set,  $X_0$  = Outcome model의 Design matrix

$X_p$  = PS model의 Design matrix 라고 보기  $\Rightarrow (1, 0, X_c^T) = X_0$  라 보기

$$\frac{\partial}{\partial \theta} \psi_{\tau}(\hat{\theta}) = \begin{bmatrix} \frac{\partial}{\partial v_1} \psi_{v_1}, \frac{\partial}{\partial v_0} \psi_{v_1}, \frac{\partial}{\partial \alpha} \psi_{v_1}, \frac{\partial}{\partial \beta} \psi_{v_1} & \text{--- (1)} \\ \frac{\partial}{\partial v_1} \psi_{v_0}, \frac{\partial}{\partial v_0} \psi_{v_0}, \frac{\partial}{\partial \alpha} \psi_{v_0}, \frac{\partial}{\partial \beta} \psi_{v_0} & \text{--- (2)} \\ \frac{\partial}{\partial v_1} \psi_{\alpha}, \frac{\partial}{\partial v_0} \psi_{\alpha}, \frac{\partial}{\partial \alpha} \psi_{\alpha}, \frac{\partial}{\partial \beta} \psi_{\alpha} & \text{--- (3)} \\ \frac{\partial}{\partial v_1} \psi_{\beta}, \frac{\partial}{\partial v_0} \psi_{\beta}, \frac{\partial}{\partial \alpha} \psi_{\beta}, \frac{\partial}{\partial \beta} \psi_{\beta} & \text{--- (4)} \end{bmatrix}$$

$$1) \psi_{v_1} = v_1 - \frac{N_1}{N} \times Y_{\tau} \times E_{\tau}$$

$$\therefore \frac{\partial}{\partial v_1} \psi_{v_1} = 1 \quad / \quad \frac{\partial}{\partial v_0} \psi_{v_1} = 0 \quad / \quad \frac{\partial}{\partial \alpha^T} \psi_{v_1} = \underbrace{0^T}_{\text{Outcome model의 Parameter 개수}} \quad / \quad \frac{\partial}{\partial \beta^T} \psi_{v_1} = \underbrace{0^T}_{\text{Propensity score model의 parameter 개수}}$$

$$(1) = (0, 0, 0^T, 0^T)$$

$$2) \psi_{v_0} = v_0 - \frac{N_1}{N} \times \left\{ Y_T \cdot (1 - E_T) \cdot \left( \frac{e^{x_{p,T}^T \beta}}{1 + e^{x_{p,T}^T \beta}} \right) + X_{-0}^T \cdot \alpha \cdot \left( E_T - \frac{e^{x_{p,T}^T \beta}}{1 + e^{x_{p,T}^T \beta}} \right) \right\} \\ \frac{1}{1 + e^{x_{p,T}^T \beta}}$$

$$= v_0 - \frac{N_1}{N} \times (1 + e^{x_{p,T}^T \beta}) \left\{ Y_T (1 - E_T) \cdot \frac{e^{\beta^T x_{p,T}}}{1 + e^{\beta^T x_{p,T}}} + X_{-0}^T \cdot \alpha \cdot \left( E_T - \frac{e^{\beta^T x_{p,T}}}{1 + e^{\beta^T x_{p,T}}} \right) \right\}$$

$$= v_0 - \frac{N_1}{N} \left\{ Y_T (1 - E_T) \cdot e^{\beta^T x_{p,T}} + (1 + e^{\beta^T x_{p,T}}) X_{-0}^T \alpha \cdot E_T - (X_{-0}^T \alpha) \cdot e^{\beta^T x_{p,T}} \right\}$$

$$= v_0 - \frac{N_1}{N} \times Y_T (1 - E_T) \cdot e^{\beta^T x_{p,T}} - \frac{N_1}{N} \times (1 + e^{\beta^T x_{p,T}}) X_{-0}^T \alpha \cdot E_T + \frac{N_1}{N} (X_{-0}^T \alpha) \cdot e^{\beta^T x_{p,T}}$$

$$\therefore \frac{\partial}{\partial v_1} \psi_{v_0} = 0, \quad \frac{\partial}{\partial v_0} \psi_{v_0} = 1, \quad \frac{\partial}{\partial \alpha^T} \psi_{v_0} = -\frac{N_1}{N} \times (1 + e^{\beta^T x_{p,T}}) \cdot E_T \cdot (X_{-0})^T$$

$$+ \frac{N_1}{N} \times e^{\beta^T x_{p,T}} \times (X_{-0})^T = -\frac{N_1}{N} \left\{ (1 + e^{\beta^T x_{p,T}}) \cdot E_T - e^{\beta^T x_{p,T}} \right\} (X_{-0})^T$$

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

$$= e^{x_{p,T}^T \beta} \left\{ -Y_T (1 - E_T) - (X_{-0}^T \alpha) \cdot E_T + (X_{-0}^T \alpha) \right\} \cdot x_{p,T}^T$$

$$= e^{x_{p,T}^T \beta} \left\{ (1 - E_T) (-Y_T + (X_{-0}^T \alpha)) \right\} x_{p,T}^T$$

$$(2) = (0, 0, \begin{bmatrix} 1 & 0 & x_c^T \end{bmatrix}, \begin{bmatrix} 1 & x_c^T \end{bmatrix})$$

$$3) \psi_\alpha = X_{0,T} (Y_T - X_{0,T}^T \alpha)$$

$$\therefore \frac{\partial}{\partial v_1} \psi_\alpha = 0, \quad \frac{\partial}{\partial v_0} \psi_\alpha = 0, \quad \frac{\partial}{\partial \alpha^T} \psi_\alpha = -X_{-0} \cdot X_{-0}^T = - \begin{bmatrix} 1 \\ 0 \\ x_c \end{bmatrix} \begin{bmatrix} 1 & 0 & x_c^T \end{bmatrix}$$

$$\frac{\partial}{\partial \beta^T} \psi_\alpha = \square \quad \begin{matrix} \text{Zero matrix} \\ (\text{앞 개수} \times \text{뒤 개수}) \end{matrix} \quad / (3) = \left[ 0, 0, - \begin{bmatrix} 1 \\ 0 \\ x_c \end{bmatrix} \begin{bmatrix} 1 & 0 & x_c^T \end{bmatrix}, \begin{bmatrix} 0 \end{bmatrix} \right]$$

$$4) \psi_{\beta} = \left( E_T - \frac{e^{x_{p,T}^T \beta}}{1 + e^{x_{p,T}^T \beta}} \right) \cdot x_{p,T}$$

$$\therefore \frac{\partial}{\partial v_1} \psi_{\beta} = \underline{0}, \quad \frac{\partial}{\partial v_0} \psi_{\beta} = \underline{0}, \quad \frac{\partial}{\partial \alpha^T} \psi_{\beta} = \boxed{0} \quad (\text{2 개수} \times \text{2 개수})$$

PS model parameter  
개수

$$\begin{aligned} \frac{\partial}{\partial \beta^T} \psi_{\beta} &= - \frac{e^{x_{p,T}^T \beta} \cdot (1 + e^{x_{p,T}^T \beta}) \cdot x_{p,T} - e^{x_{p,T}^T \beta} \cdot e^{x_{p,T}^T \beta} \cdot x_{p,T}}{(1 + e^{x_{p,T}^T \beta})^2} \cdot x_{p,T}^T = \frac{-e^{x_{p,T}^T \beta}}{(1 + e^{x_{p,T}^T \beta})^2} \cdot x_{p,T} \cdot x_{p,T}^T \\ &= - \frac{e^{x_{p,T}^T \beta}}{(1 + e^{x_{p,T}^T \beta})^2} \cdot \begin{bmatrix} 1 \\ x_c \end{bmatrix} \begin{bmatrix} 1 & x_c^T \end{bmatrix} \end{aligned}$$

$$(4) = \left[ \underline{0}, \underline{0}, \boxed{0}, - \frac{e^{x_{p,T}^T \beta}}{(1 + e^{x_{p,T}^T \beta})^2} \cdot \begin{bmatrix} 1 \\ x_c \end{bmatrix} \begin{bmatrix} 1 & x_c^T \end{bmatrix} \right]$$