

Solutions to Exercises of Chapter 12

October 26, 2025
Ting-Chih Hung
nuit0815@gmail.com

Problem 12.1 (A sanity check)

Problem 12.2 (An alternative form of the doubly robust estimator for τ)

Motivated by

$$\mu_1 = E \left\{ \frac{Z\{Y - \mu_1(X, \beta_1)\}}{e(X)} \right\} + E \{\mu_1(X, \beta_1)\},$$

we have an alternative form of the doubly robust estimator for $\mu_1 = E\{Y(1)\}$ as

$$\tilde{\mu}_1^{\text{dr2}} = \frac{E \left\{ \frac{Z\{Y - \mu_1(X, \beta_1)\}}{e(X, \alpha)} \right\}}{E \left\{ \frac{Z}{e(X, \alpha)} \right\}} + E \{\mu_1(X, \beta_1)\}.$$

We are asked to show that

$$\tilde{\mu}_1^{\text{dr2}} = \mu_1$$

if either $e(X, \alpha) = e(X)$ or $\mu_1(X, \beta_1) = \mu_1(X)$.

Let

$$L = E \left\{ \frac{Z}{e(X, \alpha)} \right\}.$$

Note that

$$\begin{aligned} & L\tilde{\mu}_1^{\text{dr2}} - L\mu_1 \\ &= E \left\{ \frac{Z\{Y - \mu_1(X, \beta_1)\}}{e(X, \alpha)} \right\} + L E \{\mu_1(X, \beta_1)\} - L\mu_1 && \text{(Definition)} \\ &= E \left\{ \frac{Z\{Y - \mu_1(X, \beta_1)\}}{e(X, \alpha)} - L\{Y(1) - \mu_1(X, \beta_1)\} \right\} && (\mu_1 = E[Y(1)]) \\ &= E \left\{ \frac{Z\{Y(1) - \mu_1(X, \beta_1)\}}{e(X, \alpha)} - L\{Y(1) - \mu_1(X, \beta_1)\} \right\} && \text{(Consistency)} \\ &= E \left\{ \frac{Z - Le(X, \alpha)}{e(X, \alpha)} \{Y(1) - \mu_1(X, \beta_1)\} \right\} && \text{(Combine terms)} \\ &= E \left\{ E \left\{ \frac{Z - Le(X, \alpha)}{e(X, \alpha)} \{Y(1) - \mu_1(X, \beta_1)\} \mid X \right\} \right\} && \text{(Iterated expectation)} \\ &= E \left\{ \frac{E \{Z - Le(X, \alpha) \mid X\}}{e(X, \alpha)} E \{Y(1) - \mu_1(X, \beta_1) \mid X\} \right\} && \text{(Ignorability)} \\ &= E \left\{ \left\{ \frac{e(X)}{e(X, \alpha)} - \frac{Le(X, \alpha)}{e(X, \alpha)} \right\} \{\mu_1(X) - \mu_1(X, \beta_1)\} \right\}. && \text{(Linearity of expectation)} \end{aligned}$$

$$= \mathbb{E} \left\{ \left\{ \frac{e(X)}{e(X, \alpha)} - L \right\} \{ \mu_1(X) - \mu_1(X, \beta_1) \} \right\}. \quad (\text{Simplification})$$

Therefore, if $e(X, \alpha) = e(X)$, then

$$L = \mathbb{E} \left\{ \frac{Z}{e(X, \alpha)} \right\} = \mathbb{E} \left\{ \frac{Z}{e(X)} \right\} = 1,$$

and thus the last line becomes

$$\mathbb{E} \{ \{1 - 1\} \{ \mu_1(X) - \mu_1(X, \beta_1) \} \} = 0.$$

On the other hand, if $\mu_1(X, \beta_1) = \mu_1(X)$, then the last line becomes

$$\mathbb{E} \left\{ \left\{ \frac{e(X)}{e(X, \alpha)} - L \right\} \{0\} \right\} = 0.$$

In both cases, we have shown that $L\tilde{\mu}_1^{\text{dr2}} - L\mu_1 = 0$, which implies that $\tilde{\mu}_1^{\text{dr2}} = \mu_1$. Similarly, we can define

$$\tilde{\mu}_0^{\text{dr2}} = \frac{\mathbb{E} \left\{ \frac{(1-Z)\{Y - \mu_0(X, \beta_0)\}}{1 - e(X, \alpha)} \right\}}{\mathbb{E} \left\{ \frac{1-Z}{1 - e(X, \alpha)} \right\}} + \mathbb{E} \{ \mu_0(X, \beta_0) \}$$

and show that

$$\tilde{\mu}_0^{\text{dr2}} = \mu_0$$

if either $e(X, \alpha) = e(X)$ or $\mu_0(X, \beta_0) = \mu_0(X)$.

The sample version of $\tilde{\mu}_1^{\text{dr2}}$ and $\tilde{\mu}_0^{\text{dr2}}$ give us an alternative form of the doubly robust estimator for τ as

$$\begin{aligned} & \hat{\tau}^{\text{dr2}} \\ &= \frac{\frac{1}{n} \sum_{i=1}^n \frac{Z_i \{Y_i - \mu_1(X_i, \hat{\beta}_1)\}}{e(X_i, \hat{\alpha})}}{\frac{1}{n} \sum_{i=1}^n \frac{Z_i}{e(X_i, \hat{\alpha})}} + \frac{1}{n} \sum_{i=1}^n \mu_1(X_i, \hat{\beta}_1) - \frac{\frac{1}{n} \sum_{i=1}^n \frac{(1-Z_i) \{Y_i - \mu_0(X_i, \hat{\beta}_0)\}}{1 - e(X_i, \hat{\alpha})}}{\frac{1}{n} \sum_{i=1}^n \frac{1-Z_i}{1 - e(X_i, \hat{\alpha})}} - \frac{1}{n} \sum_{i=1}^n \mu_0(X_i, \hat{\beta}_0). \end{aligned}$$