

**Details for the proof of Theorem 5.1:**

To see that the expectation of  $\psi(Y, W, X) \cdot \mathcal{S}(Y, W, X|\theta_0) = \frac{\partial \tau_{P,\lambda}(\theta_0)}{\partial \theta}$  we write out all sixteen components of this product ( $\psi(Y, W, X)$  has four terms, and  $\mathcal{S}(Y, W, X|\theta_0)$  has four components), and then consider the expectations of all sixteen terms separately. First,

$$\begin{aligned} & \psi(Y, W, X) \cdot s(Y, W, X|\theta_0) \\ &= \frac{W^2 \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)} \mathcal{S}_1(Y, X)(Y - \mathbb{E}[Y(1)|X]) \end{aligned} \quad (\text{C.1})$$

$$+ 0 \quad (\text{C.2})$$

$$+ \frac{W \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)} \mathcal{S}_x(X)(Y - \mathbb{E}[Y(1)|X]) \quad (\text{C.3})$$

$$+ \frac{W(W - e(X)) \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)^2(1 - e(X))} e'(X)(Y - \mathbb{E}[Y(1)|X]) \quad (\text{C.4})$$

$$- 0 \quad (\text{C.5})$$

$$- \frac{(1 - W)^2 \cdot \lambda(e(X))}{\mu_\lambda \cdot (1 - e(X))} \mathcal{S}_0(Y, X)(Y - \mathbb{E}[Y(0)|X]) \quad (\text{C.6})$$

$$- \frac{(1 - W) \cdot \lambda(e(X))}{\mu_\lambda \cdot (1 - e(X))} \mathcal{S}_x(X)(Y - \mathbb{E}[Y(0)|X]) \quad (\text{C.7})$$

$$- \frac{(1 - W) \cdot (W - e(X)) \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)(1 - e(X))^2} e'(X)(Y - \mathbb{E}[Y(0)|X]) \quad (\text{C.8})$$

$$+ \frac{W \cdot \lambda(e(X))}{\mu_\lambda} \mathcal{S}_1(Y, X)(\tau(X) - \tau_{P,\lambda}) \quad (\text{C.9})$$

$$+ \frac{(1 - W) \cdot \lambda(e(X))}{\mu_\lambda} \mathcal{S}_0(Y, X)(\tau(X) - \tau_{P,\lambda}) \quad (\text{C.10})$$

$$+ \frac{\lambda(e(X))}{\mu_\lambda} \mathcal{S}_x(X)(\tau(X) - \tau_{P,\lambda}) \quad (\text{C.11})$$

$$+ \frac{(W - e(X)) \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)(1 - e(X))} e'(X)(\tau(X) - \tau_{P,\lambda}) \quad (\text{C.12})$$

$$+ \frac{W(W - e(X)) \cdot \lambda'(e(x))}{\mu_\lambda} \mathcal{S}_1(Y, X)(\tau(X) - \tau_{P,\lambda}) \quad (\text{C.13})$$

$$+ \frac{(1 - W)(W - e(X)) \cdot \lambda'(e(x))}{\mu_\lambda} \mathcal{S}_0(Y, X)(\tau(X) - \tau_{P,\lambda}) \quad (\text{C.14})$$

$$+ \frac{(W - e(X)) \cdot \lambda'(e(x))}{\mu_\lambda} \mathcal{S}_x(X)(\tau(X) - \tau_{P,\lambda}) \quad (\text{C.15})$$

$$+ \frac{(W - e(X))^2 \cdot \lambda'(e(x))}{\mu_\lambda \cdot e(X)(1 - e(X))} e'(X)(\tau(X) - \tau_{P,\lambda}). \quad (\text{C.16})$$

Consider each expectation in turn. Equation (C.1) yields,

$$\begin{aligned}
& \mathbb{E} \left[ \frac{W^2 \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)} \mathcal{S}_1(Y, X)(Y - \mathbb{E}[Y(1)|X]) \right] \\
&= \mathbb{E} \left[ \frac{W \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)} \mathcal{S}_1(Y(1), X)Y(1) \right] - \mathbb{E} \left[ \frac{W \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)} \mathcal{S}_1(Y(1), X)\mathbb{E}[Y(1)|X] \right] \\
&= \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda \cdot e(X)} \mathbb{E}[W \cdot \mathcal{S}_1(Y(1), X)Y(1)|X] \right] \\
&\quad - \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda \cdot e(X)} \mathbb{E}[W \cdot \mathcal{S}_1(Y(1), X)|X] \cdot \mathbb{E}[Y(1)|X] \right] \\
&= \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathbb{E}[\mathcal{S}_1(Y(1), X)Y(1)|X] \right] - \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathbb{E}[\mathcal{S}_1(Y(1), X)|X] \cdot \mathbb{E}[Y(1)|X] \right] \\
&= \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathbb{E}[\mathcal{S}_1(Y(1), X)Y(1)|X] \right] \\
&= \frac{1}{\mu_\lambda} \iint \lambda(e(x))y\mathcal{S}_1(y, x)f_1(y|x)f(x)dydx.
\end{aligned}$$

Equation (C.3) yields,

$$\begin{aligned}
& \mathbb{E} \left[ \frac{W \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)} \mathcal{S}_x(X)(Y - \mathbb{E}[Y(1)|X]) \right] \\
&= \mathbb{E} \left[ \frac{W \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)} \mathcal{S}_x(X)Y(1) \right] - \mathbb{E} \left[ \frac{W \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)} \mathcal{S}_x(X)\mathbb{E}[Y(1)|X] \right] \\
&= \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda \cdot e(X)} \mathcal{S}_x(X)\mathbb{E}[WY(1)|X] \right] - \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathcal{S}_x(X)\mathbb{E}[Y(1)|X] \right] \\
&= \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathcal{S}_x(X)\mathbb{E}[Y(1)|X] \right] - \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathcal{S}_x(X)\mathbb{E}[Y(1)|X] \right] \\
&= 0.
\end{aligned}$$

Equation (C.4) yields,

$$\begin{aligned}
& \mathbb{E} \left[ \frac{W(W - e(X)) \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)^2(1 - e(X))} e'(X)(Y - \mathbb{E}[Y(1)|X]) \right] \\
&= \mathbb{E} \left[ \frac{W(W - e(X)) \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)^2(1 - e(X))} e'(X)(Y(1) - \mathbb{E}[Y(1)|X]) \right] \\
&= \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda \cdot e(X)^2(1 - e(X))} e'(X)\mathbb{E}[W(W - e(X))Y(1)|X] \right] \\
&\quad - \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda \cdot e(X)^2(1 - e(X))} e'(X)\mathbb{E}[W(W - e(X))|X] \cdot \mathbb{E}[Y(1)|X] \right] \\
&= \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda \cdot e(X)} e'(X)\mathbb{E}[Y(1)|X] \right] - \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda \cdot e(X)} e'(X)\mathbb{E}[Y(1)|X] \right] \\
&= 0.
\end{aligned}$$

Equation (C.6) yields,

$$\begin{aligned}
& -\mathbb{E} \left[ \frac{(1-W)^2 \cdot \lambda(e(X))}{\mu_\lambda \cdot (1-e(X))} \mathcal{S}_0(Y, X)(Y - \mathbb{E}[Y(0)|X]) \right] \\
&= -\mathbb{E} \left[ \frac{(1-W) \cdot \lambda(e(X))}{\mu_\lambda \cdot (1-e(X))} \mathcal{S}_0(Y(0), X)Y(0) \right] \\
&\quad + \mathbb{E} \left[ \frac{(1-W) \cdot \lambda(e(X))}{\mu_\lambda \cdot (1-e(X))} \mathcal{S}_0(Y(0), X)\mathbb{E}[Y(0)|X] \right] \\
&= -\mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda \cdot (1-e(X))} \mathbb{E}[(1-W)\mathcal{S}_0(Y(0), X)Y(0)|X] \right] \\
&\quad + \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda \cdot (1-e(X))} \mathbb{E}[(1-W)\mathcal{S}_0(Y(0), X)|X]\mathbb{E}[Y(0)|X] \right] \\
&= -\mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathbb{E}[\mathcal{S}_0(Y(0), X)Y(0)|X] \right] + \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathbb{E}[\mathcal{S}_0(Y(0), X)|X]\mathbb{E}[Y(0)|X] \right] \\
&= -\mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathbb{E}[\mathcal{S}_0(Y(0), X)Y(0)|X] \right] \\
&= -\frac{1}{\mu_\lambda} \iint \lambda(e(x))y\mathcal{S}_0(y, x)f_0(y|x)f(x)dydx.
\end{aligned}$$

Equation (C.7) yields,

$$\begin{aligned}
& -\mathbb{E} \left[ \frac{(1-W) \cdot \lambda(e(X))}{\mu_\lambda \cdot (1-e(X))} \mathcal{S}_x(X)(Y - \mathbb{E}[Y(0)|X]) \right] \\
&= -\mathbb{E} \left[ \frac{(1-W) \cdot \lambda(e(X))}{\mu_\lambda \cdot (1-e(X))} \mathcal{S}_x(X)Y(0) \right] + \mathbb{E} \left[ \frac{(1-W) \cdot \lambda(e(X))}{\mu_\lambda \cdot (1-e(X))} \mathcal{S}_x(X)\mathbb{E}[Y(0)|X] \right] \\
&= -\mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda \cdot (1-e(X))} \mathcal{S}_x(X)\mathbb{E}[(1-W)Y(0)|X] \right] + \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathcal{S}_x(X)\mathbb{E}[Y(0)|X] \right] \\
&= -\mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathcal{S}_x(X)\mathbb{E}[Y(0)|X] \right] + \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathcal{S}_x(X)\mathbb{E}[Y(0)|X] \right] \\
&= 0
\end{aligned}$$

Equation (C.8) yields,

$$\begin{aligned}
& -\mathbb{E} \left[ \frac{(1-W) \cdot (W-e(X)) \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)(1-e(X))^2} e'(X)(Y - \mathbb{E}[Y(0)|X]) \right] \\
&= -\mathbb{E} \left[ \frac{(1-W) \cdot (W-e(X)) \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)(1-e(X))^2} e'(X)(Y(0) - \mathbb{E}[Y(0)|X]) \right] \\
&= -\mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda \cdot e(X)(1-e(X))^2} e'(X)\mathbb{E}[(1-W)(W-e(X))Y(0)|X] \right] \\
&\quad + \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda \cdot e(X)(1-e(X))^2} e'(X)\mathbb{E}[(1-W)(W-e(X))|X]\mathbb{E}[Y(0)|X] \right] \\
&= -\mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda \cdot (1-e(X))} e'(X)\mathbb{E}[Y(0)|X] \right] + \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda \cdot (1-e(X))} e'(X)\mathbb{E}[Y(0)|X] \right] \\
&= 0.
\end{aligned}$$

Equation (C.9) yields,

$$\begin{aligned}
& \mathbb{E} \left[ \frac{W \cdot \lambda(e(X))}{\mu_\lambda} \mathcal{S}_1(Y, X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{W \cdot \lambda(e(X))}{\mu_\lambda} \mathcal{S}_1(Y(1), X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathbb{E}[W\mathcal{S}_1(Y(1), X)|X](\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{e(X) \cdot \lambda(e(X))}{\mu_\lambda} \mathbb{E}[\mathcal{S}_1(Y(1), X)|X](\tau(X) - \tau_{P,\lambda}) \right] \\
&= 0.
\end{aligned}$$

Equation (C.10) yields,

$$\begin{aligned}
& \mathbb{E} \left[ \frac{(1-W) \cdot \lambda(e(X))}{\mu_\lambda} \mathcal{S}_0(Y, X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{(1-W) \cdot \lambda(e(X))}{\mu_\lambda} \mathcal{S}_0(Y(0), X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathbb{E}[(1-W) \mathcal{S}_0(Y(0), X)|X](\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{(1-e(X)) \cdot \lambda(e(X))}{\mu_\lambda} \mathbb{E}[\mathcal{S}_0(Y(0), X)|X](\tau(X) - \tau_{P,\lambda}) \right] \\
&= 0.
\end{aligned}$$

Equation (C.11) yields,

$$\begin{aligned}
& \mathbb{E} \left[ \frac{\lambda(e(X))}{\mu_\lambda} \mathcal{S}_x(X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \frac{1}{\mu_\lambda} \int \lambda(e(x))(\tau(x) - \tau_{P,\lambda}) \mathcal{S}_x(x) f(x) dx
\end{aligned}$$

Equation (C.12) yields,

$$\begin{aligned}
& \mathbb{E} \left[ \frac{(W - e(X)) \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)(1-e(X))} e'(X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{\mathbb{E}[(W - e(X))|X] \cdot \lambda(e(X))}{\mu_\lambda \cdot e(X)(1-e(X))} e'(X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= 0.
\end{aligned}$$

Equation (C.13) yields,

$$\begin{aligned}
& \mathbb{E} \left[ \frac{W(W - e(X)) \cdot \lambda'(e(X))}{\mu_\lambda} \mathcal{S}_1(Y, X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{W(W - e(X)) \cdot \lambda'(e(X))}{\mu_\lambda} \mathcal{S}_1(Y(1), X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{\lambda'(e(X))}{\mu_\lambda} \mathbb{E}[W(W - e(X)) \mathcal{S}_1(Y(1), X)|X](\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{e(X)(1-e(X)) \cdot \lambda'(e(X))}{\mu_\lambda} \mathbb{E}[\mathcal{S}_1(Y(1), X)|X](\tau(X) - \tau_{P,\lambda}) \right] \\
&= 0.
\end{aligned}$$

Equation (C.14) yields,

$$\begin{aligned}
& \mathbb{E} \left[ \frac{(1-W)(W - e(X)) \cdot \lambda'(e(X))}{\mu_\lambda} \mathcal{S}_0(Y, X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{(1-W)(W - e(X)) \cdot \lambda'(e(X))}{\mu_\lambda} \mathcal{S}_0(Y(0), X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{\lambda'(e(X))}{\mu_\lambda} \mathbb{E}[(1-W)(W - e(X)) \mathcal{S}_0(Y(0), X)|X](\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{e(X)(1-e(X)) \cdot \lambda'(e(X))}{\mu_\lambda} \mathbb{E}[\mathcal{S}_0(Y(0), X)|X](\tau(X) - \tau_{P,\lambda}) \right] \\
&= 0.
\end{aligned}$$

Equation (C.15) yields,

$$\begin{aligned}
& \mathbb{E} \left[ \frac{(W - e(X)) \cdot \lambda'(e(X))}{\mu_\lambda} \mathcal{S}_x(X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{\mathbb{E}[(W - e(X))|X] \cdot \lambda'(e(X))}{\mu_\lambda} \mathcal{S}_x(X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= 0.
\end{aligned}$$

Equation (C.16) yields,

$$\begin{aligned}
& \mathbb{E} \left[ \frac{(W - e(X))^2 \cdot \lambda'(e(X))}{\mu_\lambda \cdot e(X)(1 - e(X))} e'(X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{(W^2 + e(X)^2 - 2 \cdot W e(X)) \cdot \lambda'(e(X))}{\mu_\lambda \cdot e(X)(1 - e(X))} e'(X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{\mathbb{E}[W + e(X)^2 - 2 \cdot W e(X) | X] \cdot \lambda'(e(X))}{\mu_\lambda \cdot e(X)(1 - e(X))} e'(X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{e(X)(1 - e(X)) \cdot \lambda'(e(X))}{\mu_\lambda \cdot e(X)(1 - e(X))} e'(X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \mathbb{E} \left[ \frac{\lambda'(e(X))}{\mu_\lambda} e'(X)(\tau(X) - \tau_{P,\lambda}) \right] \\
&= \frac{1}{\mu_\lambda} \int \lambda'(e(x)) e'(x) [\tau(x) - \tau_{P,\lambda}] f(x) dx
\end{aligned}$$

□

**Lemma C.1** (ASYMPTOTIC NORMALITY)

$$\sqrt{N} \cdot (\hat{\tau}_{\lambda^*} - \tau_{S,\lambda^*}) \xrightarrow{d} \mathcal{N} \left( 0, \frac{1}{(\mathbb{E}[\lambda^*(X)])^2} \cdot \mathbb{E} \left[ \lambda^*(X)^2 \cdot \left( \frac{\sigma_1^2(X)}{e(X)} + \frac{\sigma_0^2(X)}{1 - e(X)} \right) \right] \right).$$

**Proof:** By Lemma B.7, independent sampling, and because the second moment of  $\phi(Y, W, X)$  exists, it follows that

$$\sqrt{N} \cdot (\hat{\theta} - \theta_S) \xrightarrow{d} \mathcal{N} (0, \mathbb{E} [\phi(Y, W, X)^2]).$$

In addition,  $\sum_{i=1}^N \lambda^*(X_i)/N \rightarrow \mathbb{E}[\lambda^*(X)] = \mathbb{E}[e(X) \cdot (1 - e(X))]$ , so that

$$\sqrt{N} \cdot (\hat{\tau}_{\lambda^*} - \tau_{S,\lambda^*}) = \sqrt{N} \cdot \frac{1}{\sum_{i=1}^N \lambda^*(X_i)/N} \cdot (\hat{\theta} - \theta_S) \xrightarrow{d} \mathcal{N} \left( 0, \frac{1}{(\mathbb{E}[e(X) \cdot (1 - e(X))])^2} \cdot \mathbb{E} [\phi(Y, W, X)^2] \right).$$

Next,

$$\begin{aligned}
& \mathbb{E} [\phi(Y, W, X)^2] \\
&= \mathbb{E} \left[ \lambda^*(X_i)^2 \cdot \left( \left( \frac{Y \cdot W}{e(X)} - \mu_1(X) \right) - \left( \frac{Y \cdot (1 - W)}{1 - e(X)} - \mu_0(X) \right) - \left( \frac{\mu_1(X)}{e(X)} + \frac{\mu_0(X)}{1 - e(X)} \right) \cdot (W - e(X)) \right)^2 \right] \\
&= \mathbb{E} \left[ (e(X) \cdot (1 - e(X)))^2 \cdot \left( \frac{Y \cdot W}{e(X)} - \mu_1(X) \right)^2 \right] \\
&\quad - 2 \cdot \mathbb{E} \left[ (e(X) \cdot (1 - e(X)))^2 \cdot \left( \frac{Y \cdot W}{e(X)} - \mu_1(X) \right) \cdot \left( \frac{Y \cdot (1 - W)}{1 - e(X)} - \mu_0(X) \right) \right] \\
&\quad - 2 \cdot \mathbb{E} \left[ (e(X) \cdot (1 - e(X)))^2 \cdot \left( \frac{Y \cdot W}{e(X)} - \mu_1(X) \right) \cdot \left( \frac{\mu_1(X)}{e(X)} + \frac{\mu_0(X)}{1 - e(X)} \right) \cdot (W - e(X)) \right] \\
&\quad + \mathbb{E} \left[ (e(X) \cdot (1 - e(X)))^2 \cdot \left( \frac{Y \cdot (1 - W)}{1 - e(X)} - \mu_0(X) \right)^2 \right] \\
&\quad + 2 \cdot \mathbb{E} \left[ (e(X) \cdot (1 - e(X)))^2 \cdot \left( \frac{Y \cdot (1 - W)}{1 - e(X)} - \mu_0(X) \right) \cdot \left( \frac{\mu_1(X)}{e(X)} + \frac{\mu_0(X)}{1 - e(X)} \right) \cdot (W - e(X)) \right] \\
&\quad + \mathbb{E} \left[ (e(X) \cdot (1 - e(X)))^2 \cdot \left( \left( \frac{\mu_1(X)}{e(X)} + \frac{\mu_0(X)}{1 - e(X)} \right) \cdot (W - e(X)) \right)^2 \right] \\
&= \mathbb{E} \left[ (e(X) \cdot (1 - e(X)))^2 \cdot \left( \frac{\sigma_1^2(X)}{e(X)} + \mu_1^2(X) \cdot \frac{1 - e(X)}{e(X)} \right) \right] \\
&\quad + \mathbb{E} [(e(X) \cdot (1 - e(X)))^2 \cdot 2 \cdot \mu_0(X) \cdot \mu_1(X)]
\end{aligned}$$

$$\begin{aligned}
& -\mathbb{E} \left[ (e(X) \cdot (1 - e(X)))^2 \cdot \left( 2 \cdot \mu_0(X) \cdot \mu_1(X) + 2 \cdot \mu_1^2(X) \cdot \frac{1 - e(X)}{e(X)} \right) \right] \\
& + \mathbb{E} \left[ (e(X) \cdot (1 - e(X)))^2 \cdot \left( \frac{\sigma_0^2(X)}{1 - e(X)} + \mu_0^2(X) \cdot \frac{e(X)}{1 - e(X)} \right) \right] \\
& - \mathbb{E} \left[ (e(X) \cdot (1 - e(X)))^2 \cdot \left( 2 \cdot \mu_0^2(X) \cdot \frac{e(X)}{1 - e(X)} + 2 \cdot \mu_0(X) \cdot \mu_1(X) \right) \right] \\
& + \mathbb{E} \left[ (e(X) \cdot (1 - e(X)))^2 \cdot \left( 2 \cdot \mu_0(X) \cdot \mu_1(X) + \mu_1^2(X) \cdot \frac{1 - e(X)}{e(X)} + \mu_0^2(X) \cdot \frac{e(X)}{1 - e(X)} \right) \right] \\
& = \mathbb{E} \left[ (e(X) \cdot (1 - e(X)))^2 \cdot \left( \frac{\sigma_1^2(X)}{e(X)} + \frac{\sigma_0^2(X)}{1 - e(X)} \right) \right],
\end{aligned}$$

and the result in the Lemma follows.  $\square$

**Proof of Theorem: 6.2:** This follows from the combination of Lemmas B.6 and C.1.  $\square$

**Proof of Theorem: 6.3:** Again we apply Theorem 4.1 in Imbens and Ridder (2006), with as before the vector  $\tilde{Y}$  defined as

$$\tilde{Y} = \begin{pmatrix} Y \cdot W \\ Y \cdot (1 - W) \\ W \end{pmatrix},$$

Also define the functions  $\omega : \mathbb{X} \mapsto \mathbb{R}$ ,  $g : \mathbb{X} \mapsto \mathbb{R}^3$ , and  $m : \mathbb{R}^3 \mapsto \mathbb{R}^2$ :

$$\omega(x) = 1,$$

$$g(x) = \mathbb{E}[\tilde{Y}|X = x] = \begin{pmatrix} g_1(x) \\ g_2(x) \\ g_3(x) \end{pmatrix} = \begin{pmatrix} \mathbb{E}[Y \cdot W|X = x] \\ \mathbb{E}[Y \cdot (1 - W)|X = x] \\ \mathbb{E}[W|X = x] \end{pmatrix} = \begin{pmatrix} \mu_1(x) \cdot e(x) \\ \mu_0(x) \cdot (1 - e(x)) \\ e(x) \end{pmatrix},$$

$$m(z) = \begin{pmatrix} z_1 \cdot (1 - z_3) - z_2 \cdot z_3 \\ z_3 \cdot (1 - z_3) \end{pmatrix},$$

so that

$$\frac{\partial}{\partial z'} m(z) = \begin{pmatrix} 1 - z_3 & -z_3 & -z_1 - z_2 \\ 0 & 0 & 1 - 2z_3 \end{pmatrix}.$$

Then

$$\begin{aligned}
\theta & = \mathbb{E}[\omega(X) \cdot m(g(X))] = \begin{pmatrix} \mathbb{E} \left[ e(X) \cdot (1 - e(X)) \cdot \left( \frac{Y \cdot W}{e(X)} - \frac{Y \cdot (1 - W)}{1 - e(X)} \right) \right] \\ \mathbb{E} [e(X) \cdot (1 - e(X))] \end{pmatrix} \\
& = \begin{pmatrix} \mathbb{E} [(1 - e(X)) \cdot Y \cdot W - e(X) \cdot Y \cdot (1 - W)] \\ \mathbb{E} [e(X) \cdot (1 - e(X))] \end{pmatrix},
\end{aligned}$$

and

$$\hat{\theta} = \frac{1}{N} \sum_{i=1}^N \omega(X_i) \cdot m(\hat{g}(X_i)),$$

so that

$$\tau_{\lambda^*} = \theta_1 / \theta_2,$$

and

$$\hat{\tau}_{\lambda^*} = \hat{\theta}_1 / \hat{\theta}_2.$$

Assumptions 3.1-3.2 and 6.1-6.3 imply Assumptions 3.2, 3.3, 4.1, and 4.2 in Imbens and Ridder (2006). Then by Theorem 4.1 in Imbens and Ridder (2006) we have

$$\sqrt{N} \cdot (\hat{\theta} - \theta) = \frac{1}{\sqrt{N}} \sum_{i=1}^N \omega(X_i) \cdot \frac{\partial m}{\partial g'}(g(X_i)) (\tilde{Y} - g(X)) + \frac{1}{\sqrt{N}} \sum_{i=1}^N \omega(X_i) (m(g(X_i)) - \theta) + o_p(1).$$

Substituting in for  $\omega(\cdot)$ ,  $g(\cdot)$ ,  $\tilde{Y}$ , and  $m(\cdot)$ , leads to

$$\sqrt{N} \cdot (\hat{\theta} - \theta) = \frac{1}{\sqrt{N}} \sum_{i=1}^N \begin{pmatrix} \phi_1(Y_i, W_i, X_i) \\ \phi_2(Y_i, W_i, X_i) \end{pmatrix},$$

where

$$\begin{aligned} \phi_1(y, w, x) &= (1 - e(x))(y \cdot w - \mu_1(x) \cdot e(x)) - e(x) \cdot (y \cdot (1 - w) - \mu_0(x) \cdot (1 - e)) \\ &\quad - (w - e(x)) \cdot (e(x) \cdot \mu_1(x) + (1 - e(x)) \cdot \mu_0(x)) \\ &\quad + (1 - e(x)) \cdot \mu_1(x) \cdot e(x) - e(x) \cdot \mu_0(x) \cdot (1 - e(x)) - \theta_1, \end{aligned}$$

and

$$\phi_2(y, w, x) = (1 - 2 \cdot e(x)) \cdot (w - e(x)) + e(x) \cdot (1 - e(x)) - \theta_2.$$

Let

$$\mathbb{V} = \begin{pmatrix} \mathbb{V}_{11} & \mathbb{V}_{12} \\ \mathbb{V}_{12} & \mathbb{V}_{22} \end{pmatrix} = \begin{pmatrix} \mathbb{E} [\phi_1(Y, W, X)^2] & \mathbb{E} [\phi_1(Y, W, X) \cdot \phi_2(Y, W, X)] \\ \mathbb{E} [\phi_1(Y, W, X) \cdot \phi_2(Y, W, X)] & \mathbb{E} [\phi_2(Y, W, X)^2] \end{pmatrix}.$$

By a standard central limit theorem,

$$\sqrt{N} \cdot (\hat{\theta} - \theta) \xrightarrow{d} \mathcal{N}(0, \mathbb{V}),$$

and by the Delta method,

$$\sqrt{N} \cdot (\hat{\tau}_{\lambda^*} - \tau_{\lambda^*}) \xrightarrow{d} \mathcal{N}(0, \Omega),$$

where

$$\begin{aligned} \Omega &= \mathbb{V}_{11}/\theta_2^2 - 2\mathbb{V}_{12}\theta_1/\theta_2^3 + \mathbb{V}_{22}\theta_1^2/\theta_2^4 \\ &= \frac{\mathbb{E} \left[ g(X) \cdot \left( \frac{\sigma_1^2(X)}{e(X)} + \frac{\sigma_0^2(X)}{1-e(X)} + (\tau(X) - \tau_{P,g})^2 \right) + e(X)(1-e(X)) \cdot [h'(e(X))]^2 (\tau(X) - \tau_{P,\lambda})^2 \right]}{\mathbb{E} [g(X)]^2}. \end{aligned}$$

□

**Proof of Theorem 6.4:** First,

$$\hat{q}(\hat{\mathbb{A}}) = \frac{N_{\hat{\mathbb{A}}}}{N} = \frac{N_{\mathbb{A}^*}}{N} + \frac{N_{\hat{\mathbb{A}} - \mathbb{A}^*}}{N}.$$

The first term on the righthand side is equal to  $\sum_{i=1}^N \mathbb{1}\{X_i \in \mathbb{A}^*\}/N$ , which converges to  $q(\mathbb{A}^*)$  by a law of large numbers. The second term is  $o_p(1)$  by Lemma B.4, so that the lefthand side converges to  $q(\mathbb{A}^*)$ .

Next, by the Triangle Inequality,

$$\begin{aligned} & \left| \frac{1}{N_{\hat{\mathbb{A}}}} \sum_{i|X_i \in \hat{\mathbb{A}}} \left( \frac{\hat{\sigma}_1^2(X_i)}{\hat{e}(X_i)} + \frac{\hat{\sigma}_0^2(X_i)}{1 - \hat{e}(X_i)} \right) - \mathbb{E} \left[ \frac{\sigma_1^2(X)}{e(X)} + \frac{\sigma_0^2(X)}{1 - e(X)} \mid X \in \mathbb{A}^* \right] \right| \\ & \leq \left| \frac{1}{N_{\hat{\mathbb{A}}}} \sum_{i|X_i \in \hat{\mathbb{A}}} \left( \frac{\hat{\sigma}_1^2(X_i)}{\hat{e}(X_i)} + \frac{\hat{\sigma}_0^2(X_i)}{1 - \hat{e}(X_i)} \right) - \frac{1}{N_{\hat{\mathbb{A}}}} \sum_{i|X_i \in \hat{\mathbb{A}}} \left( \frac{\sigma_1^2(X_i)}{e(X_i)} + \frac{\sigma_0^2(X_i)}{1 - e(X_i)} \right) \right| \end{aligned} \quad (\text{C.17})$$

$$+ \left| \frac{1}{N_{\hat{\mathbb{A}}}} \sum_{i|X_i \in \hat{\mathbb{A}}} \left( \frac{\sigma_1^2(X_i)}{e(X_i)} + \frac{\sigma_0^2(X_i)}{1 - e(X_i)} \right) - \frac{1}{N_{\mathbb{A}^*}} \sum_{i|X_i \in \mathbb{A}^*} \left( \frac{\sigma_1^2(X_i)}{e(X_i)} + \frac{\sigma_0^2(X_i)}{1 - e(X_i)} \right) \right| \quad (\text{C.18})$$

$$+ \left| \frac{1}{N_{\mathbb{A}^*}} \sum_{i|X_i \in \mathbb{A}^*} \left( \frac{\sigma_1^2(X_i)}{e(X_i)} + \frac{\sigma_0^2(X_i)}{1 - e(X_i)} \right) - \mathbb{E} \left[ \frac{\sigma_1^2(X)}{e(X)} + \frac{\sigma_0^2(X)}{1 - e(X)} \mid X \in \mathbb{A}^* \right] \right| \quad (\text{C.19})$$

All three terms are  $o_p(1)$ , so the result follows. □

**Proof of Theorem: 6.5:** By Assumption 3.2  $\inf_{x \in \mathbb{X}} \lambda^*(x) = \inf_{x \in \mathbb{X}} e(x) \cdot (1 - e(x)) > 0$ . By Assumptions 6.1-6.3 it follows that  $\hat{\sigma}_w^2(x)$  and  $\hat{e}(x)$  converge uniformly to their limits  $\sigma_w^2(x)$  and  $e(x)$ . Hence

$$\begin{aligned}\hat{\mathbb{V}}_{S, \hat{\lambda}} &= \frac{1}{\left(\frac{1}{N} \sum_{i=1}^N e(X_i) \cdot (1 - e(X_i))\right)^2} \cdot \frac{1}{N} \sum_{i=1}^N (e(X_i) \cdot (1 - e(X_i)))^2 \cdot \left(\frac{\sigma_1^2(X_i)}{e(X_i)} + \frac{\sigma_0^2(X_i)}{1 - e(X_i)}\right) + o_p(1) \\ &= \frac{1}{\mathbb{E}[e(X) \cdot (1 - e(X))]^2} \cdot \mathbb{E} \left[ (e(X) \cdot (1 - e(X)))^2 \cdot \left(\frac{\sigma_1^2(X)}{e(X)} + \frac{\sigma_0^2(X)}{1 - e(X)}\right) \right] + o_p(1).\end{aligned}$$

□

**Proof of Theorem: 6.6:** By Assumption 3.2  $\inf_{x \in \mathbb{X}} \lambda^*(x) = \inf_{x \in \mathbb{X}} e(x) \cdot (1 - e(x)) > 0$ . By Assumptions 6.1-6.3 it follows that  $\hat{\sigma}_w^2(x)$ ,  $\hat{\tau}(x)$  and  $\hat{e}(x)$  converge uniformly to their limits  $\sigma_w^2(x)$ ,  $\tau(x)$ , and  $e(x)$ . In addition, by Theorem 6.3 it follows that  $\hat{\tau}_{S, \hat{\lambda}}$  converges to  $\tau_{P, \lambda^*}$ . Hence

$$\begin{aligned}\hat{\mathbb{V}}_{P, \hat{\lambda}} &= \frac{\frac{1}{N} \sum_{i=1}^N \lambda^*(X_i) \cdot \left(\frac{\sigma_1^2(X_i)}{e(X_i)} + \frac{\sigma_0^2(X_i)}{1 - e(X_i)} + (\tau(X_i) - \tau_{P, \omega})^2\right) + e(X_i)(1 - e(X_i)) \cdot [\lambda^{*'}(X_i)]^2 (\tau(X_i) - \tau_{P, \omega})^2}{\left(\frac{1}{N} \sum_{i=1}^N \lambda^*(X_i)\right)^2} + o_p(1) \\ &= \frac{\mathbb{E} \left[ \lambda^*(X) \cdot \left(\frac{\sigma_1^2(X)}{e(X)} + \frac{\sigma_0^2(X)}{1 - e(X)} + (\tau(X) - \tau_{P, \omega})^2\right) + e(X)(1 - e(X)) \cdot [\lambda^{*'}(X)]^2 (\tau(X) - \tau_{P, \omega})^2 \right]}{\mathbb{E} [\lambda(X)]^2} + o_p(1).\end{aligned}$$

□