

**Statistica Sinica Preprint No: SS-2016-0185.R1**

<b>Title</b>	A Further Study of Propensity Score Calibration in Missing Data Analysis
<b>Manuscript ID</b>	SS-2016-0185.R1
<b>URL</b>	<a href="http://www.stat.sinica.edu.tw/statistica/">http://www.stat.sinica.edu.tw/statistica/</a>
<b>DOI</b>	10.5705/ss.202016.0185
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Notice: Accepted version subject to English editing.	











is a model for  $E(Y | \mathbf{X})$  and  $\gamma$  is a finite-dimensional unknown parameter. Because  $R \perp Y | \mathbf{X}$  from the MAR mechanism,  $\gamma$  is conventionally estimated by  $\hat{\gamma}$  based on a complete-case analysis. When  $a(\gamma; \mathbf{X})$  is a reasonable model for  $E(Y | \mathbf{X})$ ,  $\hat{\mu}_{\text{aipw}}(\hat{\gamma})$  usually has better efficiency than  $\hat{\mu}_{\text{ipw}}$ . When  $a(\gamma; \mathbf{X})$  is correctly specified in that  $a(\gamma_0; \mathbf{X}) = E(Y | \mathbf{X})$  for some  $\gamma_0$ ,  $\hat{\mu}_{\text{aipw}}(\hat{\gamma})$  attains the semiparametric efficiency bound.

When  $a(\gamma; \mathbf{X})$  is incorrectly specified,  $\hat{\mu}_{\text{aipw}}(\hat{\gamma})$  can be inefficient (Chen, Leung, and Qin 2008; Rubin and van der Laan 2008). There have been many recent developments on gaining efficiency in this case. Two main gains have been achieved: intrinsic efficiency and improved efficiency. Estimators that are intrinsically efficient have influence functions of the form

$$\text{Resid} \left\{ \frac{R(Y - \mu_0)}{\pi(\mathbf{X})}, \frac{R - \pi(\mathbf{X})}{\pi(\mathbf{X})} h(\mathbf{X}) \right\}.$$

Hereafter, for any random variable  $\xi$  and finite-dimensional random vector  $\phi$  with mean zero and finite second moments,  $\text{Resid}(\xi, \phi) = E(\xi \phi^T) \{E(\phi \phi^T)\}^{-1} \phi$  denotes the residual of the projection of  $\xi$  onto  $\{\phi\}$ , the linear space spanned by components of  $\phi$ . Apparently the projection residual has the smallest variance among the class of influence functions  $\{R(Y - \mu_0)/\pi(\mathbf{X}) - c\{R - \pi(\mathbf{X})\}h(\mathbf{X})/\pi(\mathbf{X})$  with an arbitrary  $c$ . Various intrinsically efficient estimators have been proposed and studied by Tan (2006; 2008; 2010), Chen, Leung, and Qin (2008), Chan (2012) and Rotnitzky et al. (2012). Improved efficiency, on the other hand, is achieved by using estimator  $\tilde{\gamma}$  instead of  $\hat{\gamma}$ , where  $\tilde{\gamma}$  converges in probability to the minimizer of the asymptotic variance of  $\hat{\mu}_{\text{aipw}}(\gamma)$ . Estimators of  $\mu_0$  with improved efficiency have been proposed and studied by Rubin and van der Laan (2008), Tan (2008; 2010), Cao, Tsiatis, and Davidian (2009) and Rotnitzky et al. (2012). Many of these estimators are doubly robust: they are still consistent if  $\pi(\alpha; \mathbf{X})$  is misspecified but  $a(\gamma; \mathbf{X})$  is not.































































