

# MATH 4824C - Sample Final

May 15, 2024

Please (1) show the work to the questions and (2) return the answer sheet on time.

1. In randomized trials; i.e.  $(Y_i(1), Y_i(0), X_i) \perp Z_i$ , show that  $Var(Y(1)) = Var(Y|Z = 1)$ ;
2. Show that in observational studies without unmeasured confounders; i.e.  $(Y_i(1), Y_i(0)) \perp Z_i | X_i$ , show that  $\alpha_0 = E(Y(1) - X\beta_0)$  where  $(\alpha_0, \beta_0) = \arg \min_{\alpha, \beta} E[Z(Y - \alpha - X\beta)^2 / e(X)]$  and  $e(X) = P(Z = 1|X)$ ;
3. Show that in observational studies without unmeasured confounders; i.e.  $(Y_i(1), Y_i(0)) \perp Z_i | X_i$ , let  $e(x, \beta)$ , be the logistic model of  $Z_i$  on  $X_i$ , and  $\mu_1(x, \alpha_1)$  be linear or logistic model of  $Y_i(1)$  on  $X_i$ . Try to show that

$$\tilde{\mu}_{1,DR} = \frac{E(Z_i \{Y_i - \mu_1(X_i, \alpha_1)\} / e(X_i, \beta))}{E(Z_i / e(X_i, \beta))} + E(\mu_1(X_i, \alpha_1))$$

is doubly robust, that is, if either  $\mu_1(x, \alpha_1) = \mu_1(x) := E[Y_i|Z_i = 1, X_i = x]$  or  $e(x, \beta) = e(x) := P(Z_i|X_i = x)$ , then  $\tilde{\mu}_{1,DR} = E\{Y_i(1)\}$ .

4. Consider linear models  $Y_i = \beta_0 + \beta_1 D_i + X_i^\top \beta_X + \epsilon_i$  and  $D_i = \gamma_0 + \gamma_1 Z_i + X_i^\top \gamma_X + \eta_i$  where  $Z_i$  and  $X_i$  are endogeneous,  $\epsilon_i$  is independent of  $Z_i$  and  $\eta_i$  is independent of  $X_i$ , write down the reduced form of  $Y_i \sim Z_i + X_i$  and show that both  $Z$  and  $X$  are exogeneous in the reduced model.
5. Consider the true linear model  $Y = \beta_0 + \beta_1 D + \beta_2 U + \eta$  and  $D = \alpha_0 + \alpha_1 Z + \epsilon$ , show that  $\beta_1 = \frac{cov(Y, Z)}{cov(D, Z)}$  if  $cov(D, Z) \neq 0$  and  $Z$  is independent of  $U$  and  $\eta$ .
6. In regression discontinuity design with forcing variable  $X$  and treatment indicator  $Z = I_{X \geq c}$  where  $c$  is the cutoff, show that  $E(Y_i(0) | X_i = c) = \lim_{x \uparrow c} E(Y_i | X_i = x)$  if  $E(Y_i(0) | X_i = c)$  is left continuous at  $c$ .
7. In the non-compliance observational study where the exclusion and the monotonicity assumption are satisfied and  $Z_i \perp \{Y_i(1), Y_i(0), D_i(1), D_i(0)\} | \mathbf{X}_i$ , show that  $E\{Y_i(1) - Y_i(0) | complier, X = x\} = (E(Y | Z = 1, X = x) - E(Y | Z = 0, X = x)) / (E(D | Z = 1, X = x) - E(D | Z = 0, X = x))$

8. In mediation analysis where there exist no treatment-outcome confounding, no mediator-outcome confounding and no treatment-mediator confounding and the cross-world assumption is satisfied, write down the identifiable quantity of  $E(Y(0, M(1)) - E(Y(0, M(0)))$ .