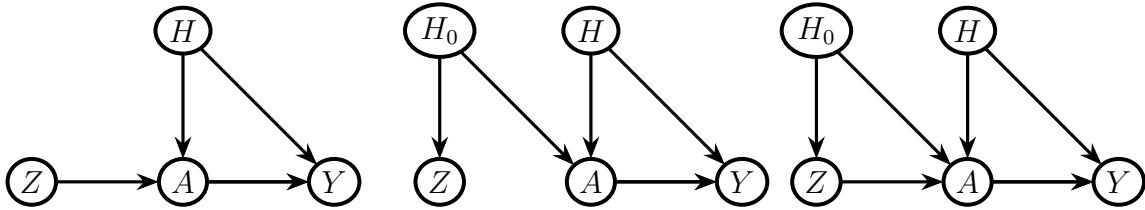


## EXERCISES FOR RANDOMIZATION AND CAUSATION (MATH-336)

### EXERCISE SHEET 11

**Exercise 1** (Instrumental variables). (From [1]) Consider an instrumental variable setting which is described by one of the following three DAGs.



(a) Can we use the main IV assumptions (1)-(3) to infer any (conditional) independencies between the observed variables  $A, Z, Y$ , that is, any factorization of the law  $p(y, a, z)$  that describes the observed data? We reproduce the main IV assumptions below for convenience:

- (1)  $\text{cor}(Z, A) \neq 0$  (instrument strength)
- (2)  $Y^{z,a} = Y^a$  for all  $a, z$  (exclusion restriction)
- (3)  $Z \perp\!\!\!\perp Y^a$  for all  $a$  (unconfoundedness of  $Z$ ).

(b) Consider the following structural equation model for  $Y$ :

$$(1) \quad Y = f_Y(A, H, \epsilon_Y) = h(\epsilon_Y)A + g(H, \epsilon_Y) .$$

The model does allow certain effect heterogeneity, because the individual level causal effect

$$Y^a - Y^{a'} = h(\epsilon_Y)(a - a')$$

is a random variable. The average causal effect is defined as

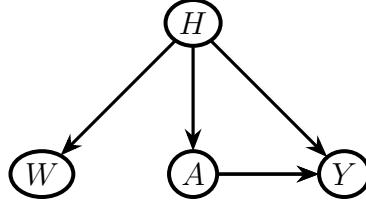
$$E[Y^a] - E[Y^{a'}] = E[h(\epsilon_Y)](a - a') .$$

Assume that the linear structural equation model Eq. 1 holds, that  $Y^{a=0} \perp\!\!\!\perp Z$  and that  $E[h(\epsilon_Y) | Z, A] = E[h(\epsilon_Y)]$ . Show that the additive average causal effect is then given by

$$E[h(\epsilon_Y)] = \frac{\text{Cov}(Z, Y)}{\text{Cov}(Z, A)} .$$

(c) Assume that the model in Eq. 1 holds, and that  $E[h(\epsilon_Y) | Z, A] = E[h(\epsilon_Y)]$ . Show that then, there exists a constant  $\beta$  such that

$$E[Y | Z, A] - E[Y^0 | Z, A] = \beta A .$$



**Exercise 2** (A sensitivity analysis). Consider the treatment  $A$ , outcome  $Y$ , unmeasured variable  $H$  and measured pre-treatment variable  $W$  satisfying the graph below.

As we can see from the graph, both  $W$  and  $A$  are confounded for  $Y$  by  $H$ . Suppose that

$$E[Y^{a=0} | A = 1] - E[Y^{a=0} | A = 0] = E[W | A = 1] - E[W | A = 0] .$$

- (a) Use this assumption to find an identification formula for  $E[Y^{a=1} - Y^{a=0} | A = 1]$  in terms of the observed data  $A, W, Y$ .
- (b) Can we interpret this as an average total effect in the entire population?

**Exercise 3** (Sensitivity analysis with IVs). Consider a binary instrument  $Z$ , a binary treatment  $A$  and a binary outcome  $Y$  satisfying:

- (1) Exclusion restriction:  $Y^{z,a} = Y^a$
- (2) IV exchangeability:  $Y^a \perp\!\!\!\perp Z$

Show that under assumptions (1)-(2),

$$P(Y = 0, A = 1 | Z = 0) + P(Y = 1, A = 1 | Z = 1) \leq 1 .$$

*Hint:* Use the fact that  $p(x_1, x_2 | x_3) \leq p(x_1 | x_3)$ . Likewise, it can also be shown that

- $P(Y = 0, A = 1 | Z = 0) + P(Y = 1, A = 1 | Z = 1) \leq 1$
- $P(Y = 0, A = 1 | Z = 1) + P(Y = 1, A = 1 | Z = 0) \leq 1$
- $P(Y = 0, A = 0 | Z = 1) + P(Y = 1, A = 0 | Z = 0) \leq 1$

These inequalities can be used to falsify IV exchangeability assumption. With some more arguments, it is also possible to use the IV inequalities to obtain bounds on causal effects.

## REFERENCES

- [1] Andrea Rotnitzky. BST 257 (Harvard T.H. Chan School of Public Health).