

METHODS FOR CAUSAL INFERENCE: TUTORIAL 4

1. [*Causal Inference in statistics: A primer*, Chapter 3]
 - (a) List all of the sets of variables that satisfy the backdoor criterions, to identify the causal effect of T on Y .
 - (b) List all of the *minimal* sets of variables that satisfy the backdoor criterion, to identify the causal effect of T on Y . A *minimal* sets of variables, refers to variables which if removed, the backdoor criterion is no longer satisfied.
 - (c) List all minimal sets of variable that need to be measured in order to indentify the effect of D on Y. Repeat for the effect of $\{W, D\}$ on Y .
 - (d) Assume only one variable, apart from T and Y, can be measure for the above graph. Explain which variable would allow the identification of the causal effect of T on Y and write down the identification formula.

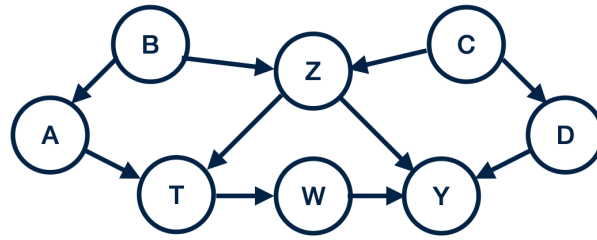


FIG. 1: The causal graph (Lecture 14)

2. Create a simulation similar to Lecture 13 and convince yourself of the optimal adjustment rule using numerical examples and causal graphs.
3. Using the lecture notes convince yourself that

$$p(Y_x = y) = p(Y = y | do(T = t)), \quad (1)$$

by considering the set Z which satisfies the backdoor criterion.

4. [*Causal Inference in statistics: A primer*, Chapter 3] Prove that if T is binary, the effect of treatment on the treated can be estimated from a mix of observational and experimental data.

Hint: Decompose $\mathbb{E}[Y_t] = \mathbb{E}[Y_t | T = t']p(T = t') + \mathbb{E}[Y_t | T = t]p(T = t)$ and use the observation in the previous question.
