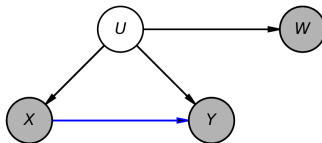


Problem Statement

Identification: Can the causal effect of X on Y be uniquely estimated given the distribution and model structure?

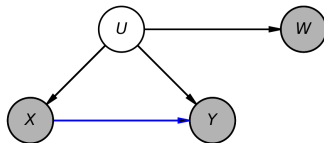


$$pr\{y|do(x)\} = \sum_u pr(y|x, u)pr(u)$$

Case 1: Single proxy variable with known error mechanism

Required: $pr(w|u)$

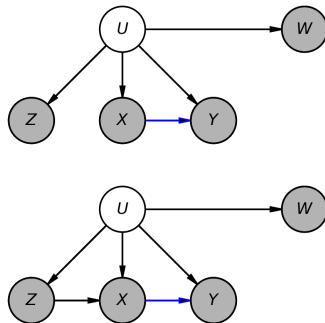
Use matrix adjustment to estimate $pr(y|x, u)$ and $pr(u)$.



Case 2: Multiple proxy variables.

Required: One proxy variable should only be connected to U .

1. Choose one of the proxy variables as the “indicator variable”, W .
2. Use eigen value decomposition to estimate $pr(w|u)$.
3. Follow the method of first case.

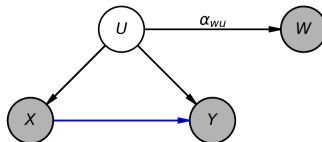


Case 3: Linear SEM

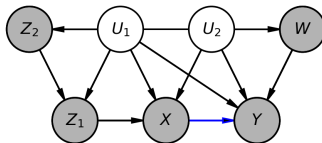
Required: The contribution of U 's variance to W 's variance:

$$\alpha_{wu}^2 \sigma_{uu} = \sigma_{ww} - \sigma_{\epsilon_w, \epsilon_w}$$

1. If single proxy variable, needs $\alpha_{wu}^2 \sigma_{uu}$.
2. If multiple proxy variables, $\alpha_{wu}^2 \sigma_{uu}$ can be computed.



Case 4: Linear SEM with instrumental variables



THEOREM 2. *Suppose that:*

- (i) *a nonempty set $\{Z_1, Z_2\}$ of distinct variables satisfies one of the following conditions: (i-a) both Z_1 and Z_2 are conditional instrumental variables given a univariate U relative to (X, Y) , (i-b) Z_1 is a conditional instrumental variable given U relative to (X, Y) , and $Z_2 = X$ and U satisfies the back door criterion relative to (X, Y) , (i-c) Z_1 is a conditional instrumental variable given U relative to (X, Y) , and U d-separates Z_2 from $\{X, Y\}$;*
- (ii) *U d-separates $\{Z_1, Z_2\}$ from an observed variable W .*

Then the total effect τ_{yx} of X on Y is identifiable and is given by the formula (11).