Counterfactuals and Mediation

Brady Neal

causalcourse.com

Counterfactuals Basics

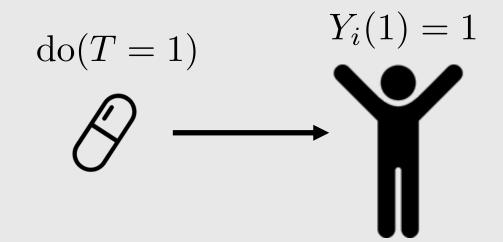
Important Application: Mediation

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Counterfactuals Basics

Important Application: Mediation

Fundamental Problem of Causal Inference



T: observed treatment

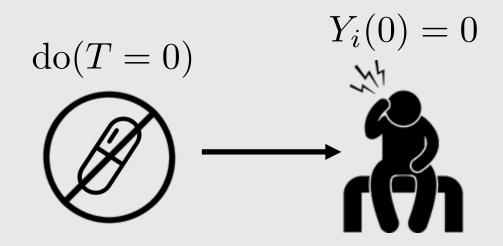
: observed outcome

i : used in subscript to denote a

specific unit/individual

 $Y_i(1)$: potential outcome under treatment

 $Y_i(0)$: potential outcome under no treatment

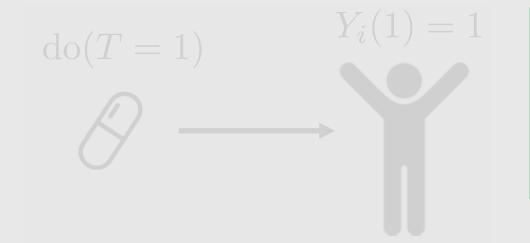


Causal effect

$$Y_i(1) - Y_i(0) = 1$$

Fundamental Problem of Causal Inference

Counterfactual



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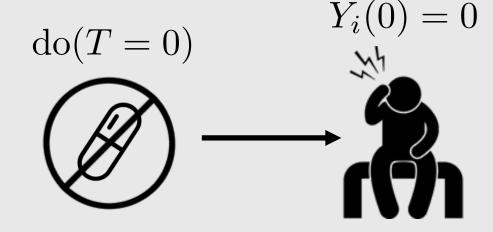
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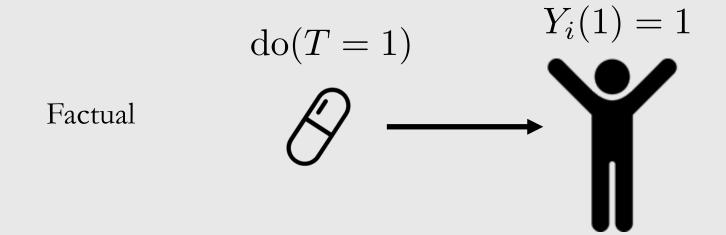
Factual



Causal effect

$$Y_i(1) - Y_i(0) = 1$$

Fundamental Problem of Causal Inference



T: observed treatment

Y: observed outcome

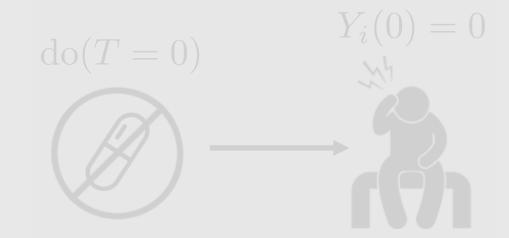
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Counterfactual



Causal effect

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We can compute counterfactuals using a parametric SCM.

Counterfactual: $P(Y(t) \mid T = t', Y = y')$

observation Counterfactual: $P(Y(t) \mid T = t', Y = y')$

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Different from CATE: $\mathbb{E}[Y(t) \mid X = x] = \mathbb{E}[Y \mid do(t), X = x]$

Cannot express counterfactuals using do-notation

Given: Observation of (T, Y) (observation of potential outcome Y(t) where t is the observed value of T)

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Main ingredient necessary: correct parametric model for the structural equation for Y

Result: access to counterfactuals Y(t') at the unit-level

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$$T := ...$$

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Observation: T = 0 and Y = 0

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Step 1: Solve for U

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Step 2: Individualized SCM

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Computing Counterfactuals: Simple Example

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ITE:
$$Y_u(1) - Y_u(0) = 1 - 0 = 1$$

From Chapter 4 of Pearl et al. (2016)'s Primer:

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Question:

Given the observation T = 1 and Y = 0, compute Y(0) for this individual given the following SCM:

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Example:

$$Y := egin{cases} 1 & U = ext{always happy} \\ 0 & U = ext{never happy} \\ T & U = ext{dog-needer} \\ 1 - T & U = ext{dog-hater} \end{cases}$$

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$$P(U = \text{always happy}) = 0.3$$

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Observation: T = 1 and Y = 0
$$(Y_u(1) = 0) \qquad P(U = \text{never happy} \mid T = 1, Y = 0) = \frac{0.2}{0.2 + 0.1} = \frac{2}{3}$$
$$P(U = \text{dog-hater} \mid T = 1, Y = 0) = \frac{0.1}{0.2 + 0.1} = \frac{1}{3}$$
$$Y_u(0) = ?$$

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Main ingredient necessary for computing counterfactuals: parametric model for the structural equation for Y

Strong assumption

Without it, we are stuck with the fundamental problem of causal inference.

Question:

Given the observation T = 1 and Y = 1, compute Y(0) for this individual given the following SCM and prior:

$$Y := \begin{cases} 1 & U = \text{always happy} \\ 0 & U = \text{never happy} \\ T & U = \text{dog-needer} \\ 1 - T & U = \text{dog-hater} \end{cases} \qquad P(U = \text{always happy}) = 0.3$$

$$P(U = \text{never happy}) = 0.2$$

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