

Causal Inference in Three-Wave Panel Designs

Joseph.Bulbulia@vuw.ac.nz // Victoria University

Section 1

The Fundamental Problem of Causal Inference

Causality Requires a Contrast of Exposures

$$Y_{\text{you}}(1) - Y_{\text{you}}(0)$$

But Individuals Experience Only One Exposure

$Y_i|A_i = 1 \implies Y_i(0)|A_i = 1$ is counterfactual

Average Treatment Effect in Randomised Controlled Experiments Work From Assumptions

$$\text{Average Treatment Effect} = \left[\begin{array}{c} \left(\underbrace{\mathbb{E}[Y(1)|A=1]}_{\text{observed}} + \underbrace{\mathbb{E}[Y(1)|A=0]}_{\text{unobserved}} \right) \\ - \left(\underbrace{\mathbb{E}[Y(0)|A=0]}_{\text{observed}} + \underbrace{\mathbb{E}[Y(0)|A=1]}_{\text{unobserved}} \right) \end{array} \right]$$

Section 2

The Three Fundamental Assumptions of Causal Inference

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Causal Consistency

$$Y_i^{observed}|A_i = \begin{cases} Y_i(a^*) & \text{if } A_i = a^* \\ Y_i(a) & \text{if } A_i = a \end{cases}$$

Conditional Exchangeability

$$Y(a) \coprod A|L \quad \text{or equivalently} \quad A \coprod Y(a)|L$$

Positivity

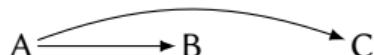
$$0 < Pr(A = a|L = l) < 1$$

Section 3

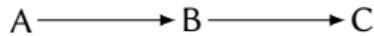
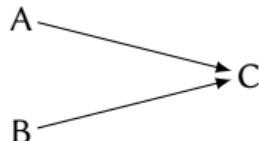
Causal Diagrams & Structures of Confounding Bias

1 **Causality Absent**

A B

A and *B* have no causal effect on each other. $A \perp\!\!\!\perp B$ (independent)2 **Causality** $A \rightarrow B$ *A* causally affects *B*, and they are associated. $A \not\perp\!\!\!\perp B$ (dependent)3 **Fork**

Three variables

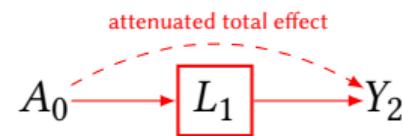
A causally affects both *B* and *C*; *B* and *C* are conditionally independent given *A*. $B \perp\!\!\!\perp C | A$ 4 **Chain***C* is affected by *B* which is, in turn, affected by *A*; *A* and *C* are conditionally independent given *B*. $B \perp\!\!\!\perp C | A$ 5 **Collider***C* is affected by both *A* and *B*, which are independent; conditioning on *C* induces association between *A* and *B*. $A \not\perp\!\!\!\perp B | C$

Confounding

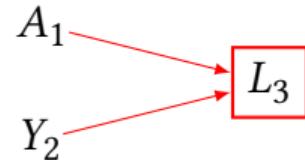
Reverse Causation

$$Y_1 \longrightarrow A_2$$

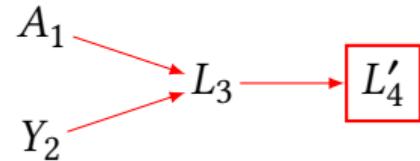
Mediator Bias



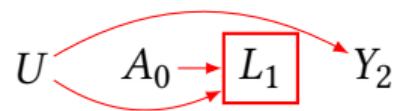
Collider Bias



Collider Bias Proxy



Post Exposure Collider Bias



Unmeasured Common Cause

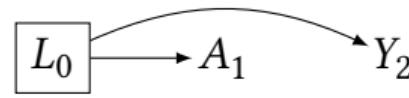
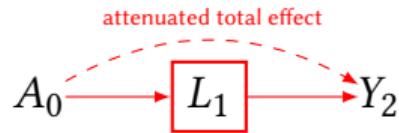


Reverse Causation Solution: Longitudinal Hygiene

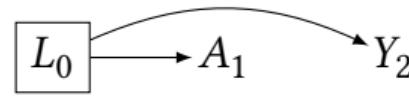
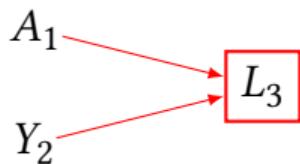
$$Y_1 \longrightarrow A_2$$

$$A_1 \qquad Y_2$$

Mediator Bias Solution: Longitudinal Hygiene

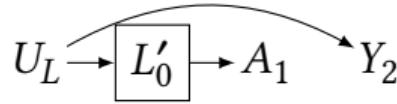
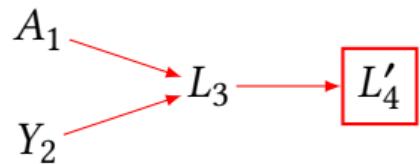


Collider Bias Solution: Longitudinal Hygiene

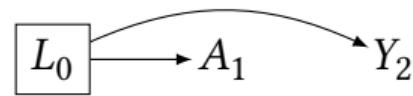
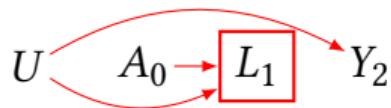


More Confounding Solutions: Longitudinal Hygiene

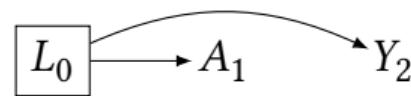
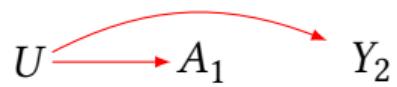
Collider Bias Proxy Solution: Longitudinal Hygiene



Post Exposure Collider Bias Solution: Longitudinal Hygiene



Unmeasured Common Cause Solution: Longitudinal Hygiene



Section 4

New Zealand Attitudes and Values Study

The New Zealand Attitudes and Values Study

- Planned 20-year longitudinal study, currently in its 14th year.

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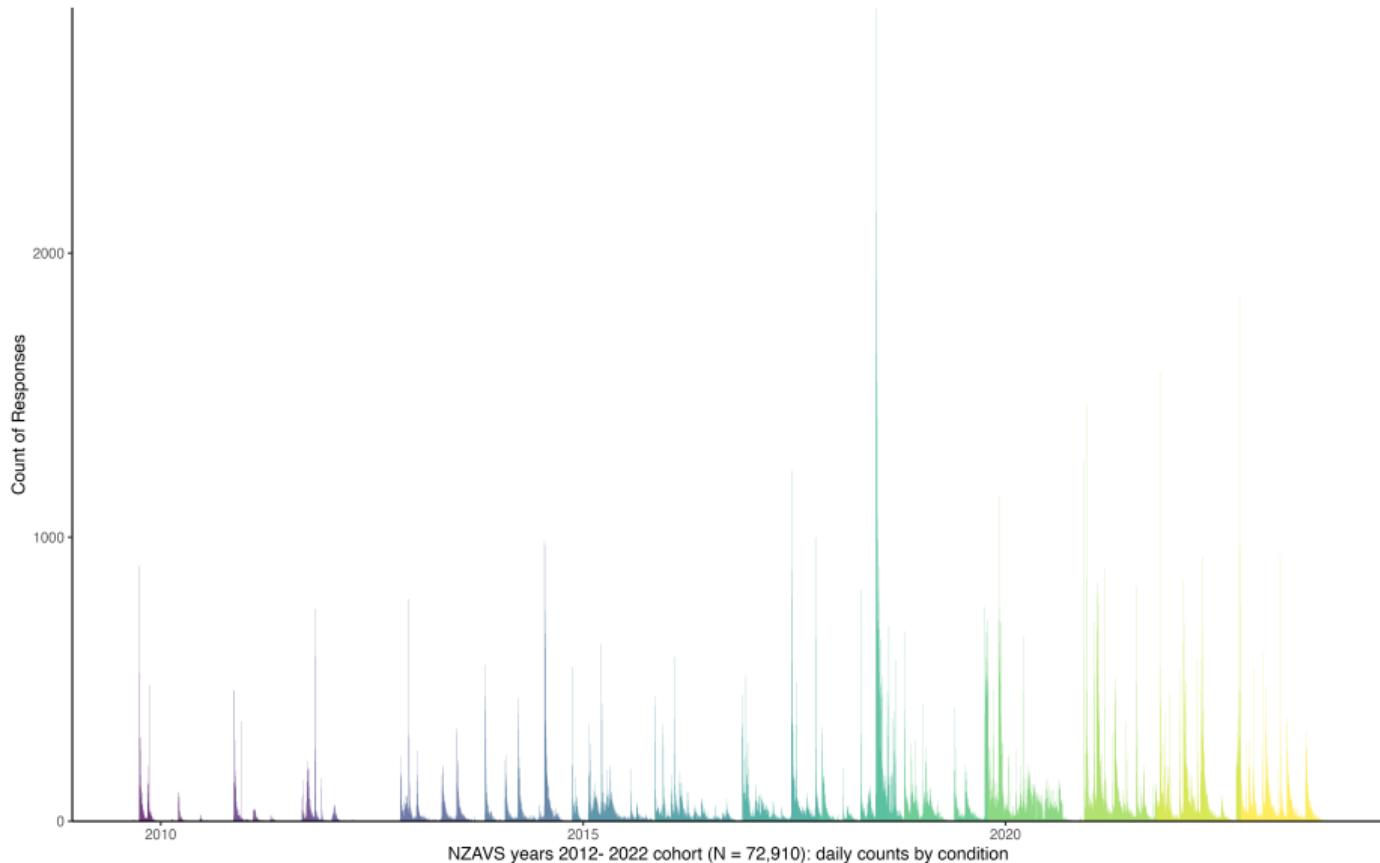
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- Postal questionnaire (coverage; retention ~ 80%)
- Large multidisciplinary research team (40 +)
- Focus on personality, social attitudes, values, religion, employment, prejudice ...
- Current sample contains > 72,290,000 unique people, and ~ 38,000 in the longitudinal study

New Zealand Attitudes and Values Study (panel)

N = 72,910; years 2012-2022

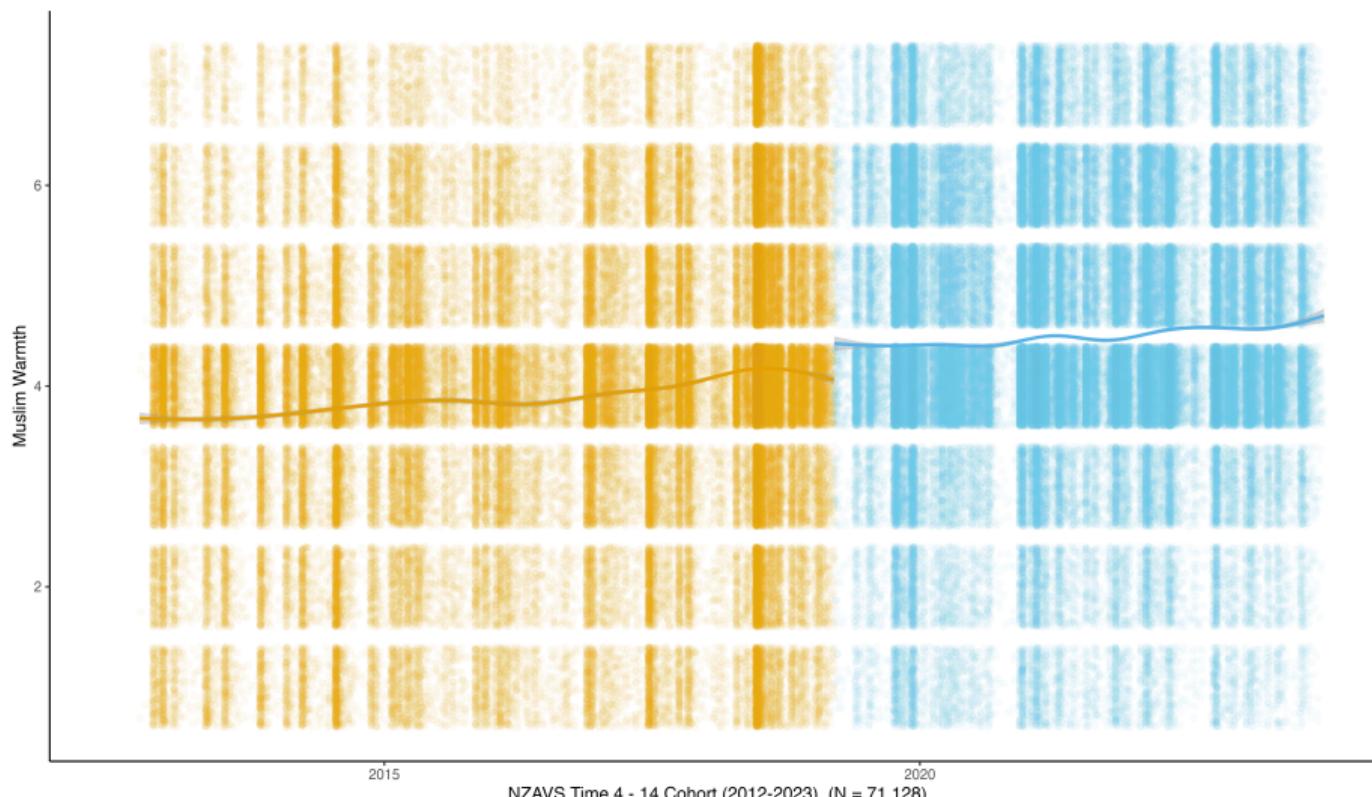


Big Events

Discontinuity at attacks (GAM)

Boost to Warmth increase in the years following the attacks: FULL SAMPLE

attack_condition 0 1



Gradual Events

Religious Service Attendance

Socialising

Cash Value of Religious Service

Thanks

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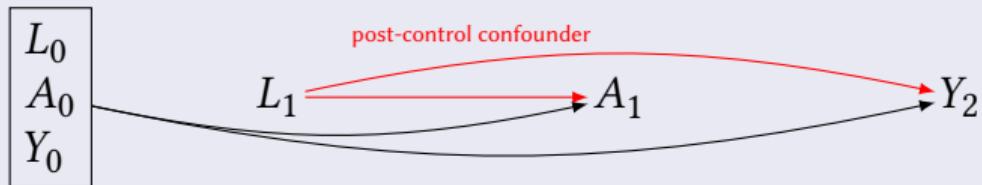
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- Max Planck Institute for Evolutionary Anthropology
- Victoria University
- University of Canterbury
- 72,290 NZAVS participants

Section 5

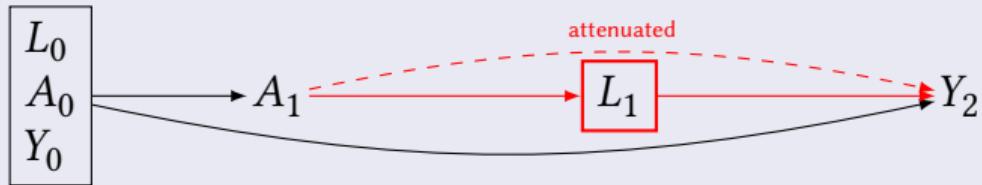
Extra Slides

Longitudinal Data Bring Their Own Problems

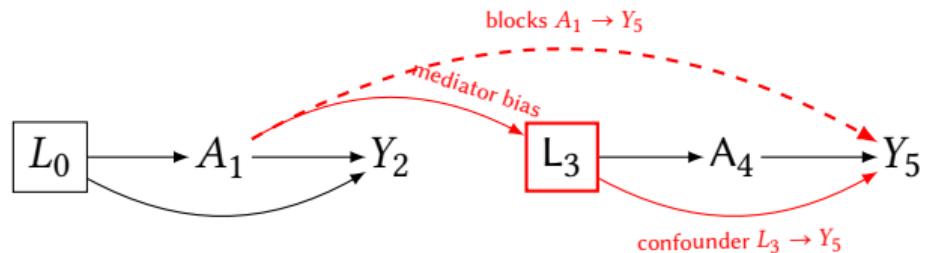
Timing of Confounder



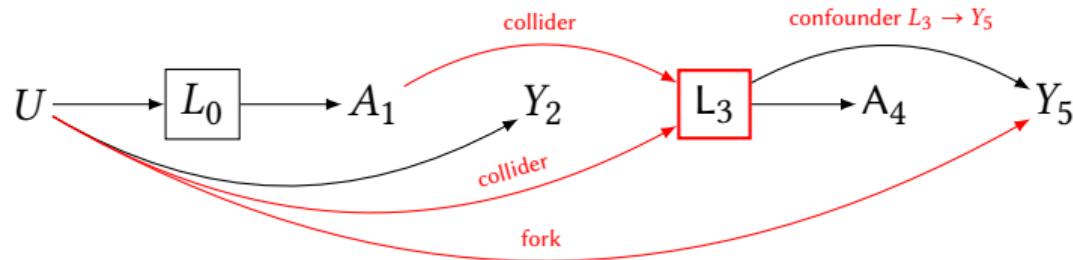
Timing of Mediator



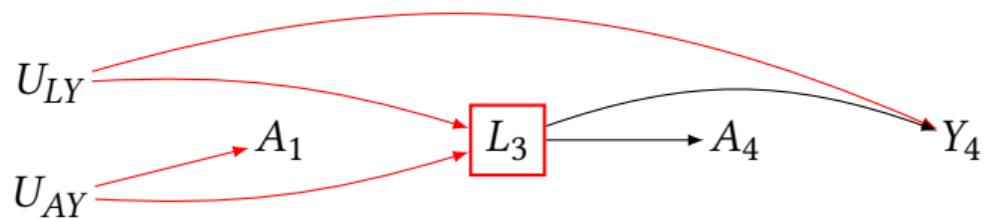
Treatment Confounder Bias



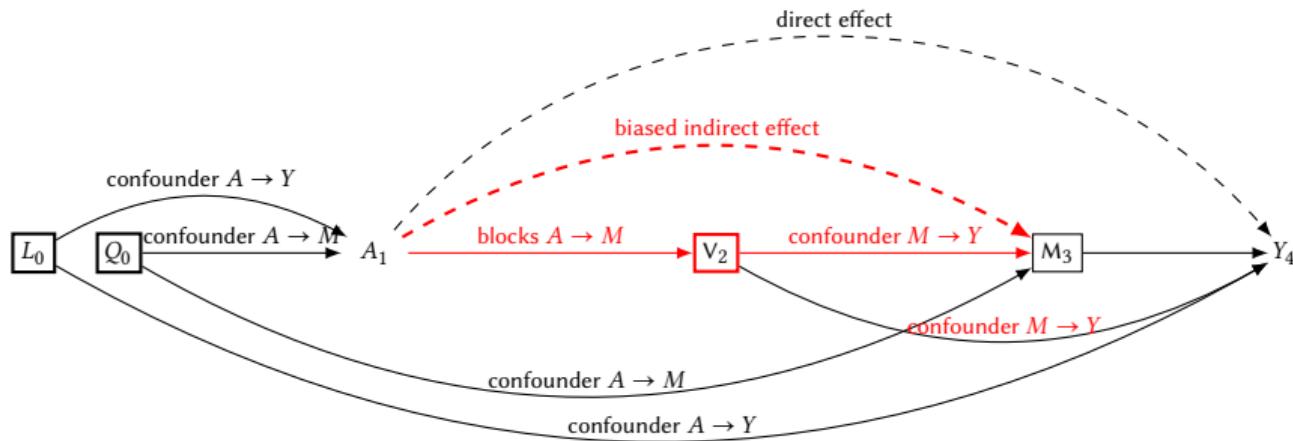
Treatment Confounder Feedback



Treatment Confounder Feedback Variation



Mediation



Section 6

Mediation

Total Effect

$$TE = \mathbb{E}[Y(1)] - \mathbb{E}[Y(0)]$$

Total Effect Considering Mediator

$$TE = \mathbb{E}[Y(1)] - \mathbb{E}[Y(0)]$$

$$\mathbb{E}[Y(1)] = \mathbb{E}[Y(1, M(1))]$$

Natural Direct Effect

Natural Direct Effect (NDE) is the effect of the treatment on the outcome while maintaining the mediator at the level it would have been if the treatment had *not* been applied:

$$NDE = \mathbb{E}[Y(1, M(0))] - \mathbb{E}[Y(0, M(0))]$$

Natural Indirect Effect

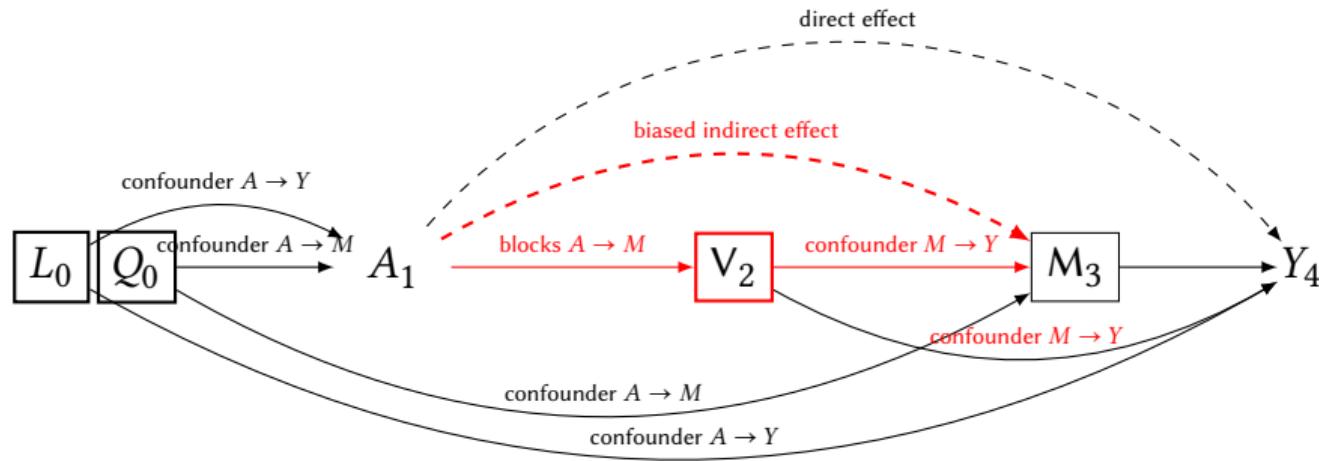
Natural Indirect Effect (NIE): is the effect of the exposure on the outcome that is mediated. To obtain these quantities we must compare the potential outcome Y under treatment, where the mediator assumes its natural level under treatment with the potential outcome when the mediator assumes its natural value under no treatment is given:

$$NIE = \mathbb{E}[Y(1, M(1))] - \mathbb{E}[Y(1, M(0))]$$

Decomposition

$$\text{Total Effect (TE)} = \underbrace{\left\{ \mathbb{E}[Y(1, M(1))] - \mathbb{E}[Y(1, M(0))] \right\}}_{\text{Natural Indirect Effect (NIE)}} + \underbrace{\left\{ \mathbb{E}[Y(1, M(0))] - \mathbb{E}[Y(0, M(0))] \right\}}_{\text{Natural Direct Effect (NDE)}}$$

Why Mediation is Difficult



Section 7

Interaction

Interaction: potential outcomes

$$\left(\underbrace{\mathbb{E}[Y(1, 1)]}_{\text{joint exposure}} - \underbrace{\mathbb{E}[Y(0, 0)]}_{\text{neither exposed}} \right) - \left[\left(\underbrace{\mathbb{E}[Y(1, 0)]}_{\text{only A exposed}} - \underbrace{\mathbb{E}[Y(0, 0)]}_{\text{neither exposed}} \right) + \left(\underbrace{\mathbb{E}[Y(0, 1)]}_{\text{only B exposed}} - \underbrace{\mathbb{E}[Y(0, 0)]}_{\text{neither exposed}} \right) \right] \neq 0$$

Interaction: simplifies to

$$\underbrace{\mathbb{E}[Y(1, 1)]}_{\text{joint exposure}} - \underbrace{\mathbb{E}[Y(1, 0)]}_{\text{only A exposed}} - \underbrace{\mathbb{E}[Y(0, 1)]}_{\text{only B exposed}} + \underbrace{\mathbb{E}[Y(0, 0)]}_{\text{neither exposed}} \neq 0$$

Key

Symbol

X	Any variable.
A	The treatment or, equivalently, the exposure.
Y	The outcome.
$Y(a)$	The potential outcome when $A = a$.
L	Measured confounder(s): typically comprises a set of variables.
U	Unmeasured confounder.
Z	Effect-modifier (or ‘moderator’) of A on Y .
M	Mediator of A on Y .
\bar{X}	Sequential variables, e.g. $\bar{A} = \{A_1, A_2, A_3\}; \bar{L} = \{L_0, L_1, L_2\}$.
\mathcal{R}	Denotes randomisation into treatment event.
