Discussion of Lok and Bosch (2020)

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Contributions

 Reinterpretation of the mediation formula based on the organic indirect and direct effects (Lok 2016)

- avoids cross-world counterfactuals
- use of stochastic intervention on mediator

In linear models, the organic indirect and direct effects under the control condition equals the product method of Baron and Kenny

- the assumption of no treatment-mediator interaction is not required
- different from the organic indirect and direct effects under the treatment condition



Natural Indirect and Direct Effects

• Natural indirect effects

$$\zeta(a) = \mathbb{E}\{Y_i(a, M_i(1)) - Y_i(a, M_i(0))\} \text{ for } a = 0, 1$$

Natural direct effects

$$\xi(a) = \mathbb{E}\{Y_i(1, M_i(a)) - Y_i(0, M_i(a))\}$$
 for $a = 0, 1$

- We must infer cross-world counterfactual $\mathbb{E}\{Y_i(a, M_i(a'))\}$ for $a \neq a'$
- Even in RCT, we need the cross-world independence assumption:

$$Y_i(a', m) \perp M_i(a) \mid A_i = a, C_i \text{ for any } a, a' \text{ and } C_i$$

I randomization of mediator does not satisfy this

2 no post-treatment confounder allowed

Organic Indirect and Direct Effects

• Organic indirect effect:

$$\zeta^*(1) = \mathbb{E}\{Y_i(1, M_i(1))\} - \sum_m \mathbb{E}\{Y_i(1, m)\} \cdot \Pr(M_i(0) = m)$$

• Organic direct effect:

$$\xi^*(0) = \sum_m \mathbb{E}\{Y_i(1,m)\} \cdot \Pr(M_i(0) = m) - \mathbb{E}\{Y_i(0,M_i(0))\}$$

- They do not require cross-world counterfactuals and assumptions
- Contrast this with the natural effects counterpart:

$$\mathbb{E}\{Y_i(1, M_i(0))\} = \sum_m \mathbb{E}\{Y_i(1, m) \mid M_i(0) = m\} \cdot \Pr(M_i(0) = m)$$

• Cross-world independence implies:

$$\mathbb{E}\{Y_i(1,m) \mid M_i(0) = m\} = \mathbb{E}\{Y_i(1,m)\}$$

Relation between Organic and Natural Effects

Organic indirect and direct effects generalize natural indirect and direct effects: provided that $M_{1,I=1} = M_0$ exists, the usual crossworlds assumption implies that the intervention I that sets the mediator to M_0 , $M_{1,I=1} = M_0$, is organic! (page 5)

- Questions:
 - Natural effects become organic effects under cross-world independence organic effects are special cases of natural effects?
 - 2 What does it mean to say $M_{1,l=1} = M_0$ exists?
 - For natural effects, the intervention is about individual values whereas for organic effects, it is about their distribution
 - This seems only possible in the limiting case where the mediator is a deterministic function of A and C → violation of positivity assumption

New Organic Indirect and Direct Effects

• Organic indirect effect:

$$\zeta^*(0) = \sum_m \mathbb{E}\{Y_i(0,m)\} \cdot \Pr(M_i(1) = m) - \mathbb{E}\{Y_i(0,M_i(0))\}$$

Organic direct effect:

$$\xi^*(0) = \mathbb{E}\{Y_i(1, M_i(1))\} - \sum_m \mathbb{E}\{Y_i(0, m)\} \cdot \Pr(M_i(1) = m)$$

With the new definition, the product method works for linear models regardless of whether the outcome model has treatmentmediator interaction. (page 7)

- This is consistent with the standard result for the natural effects
- Product method for binary mediator is also applicable to natural effects

Concluding Remarks

- Organic indirect and direct effects:
 - avoid cross-world counterfactuals
 - yet the same mediation formula and all the standard results apply
 - different interpretation and same methods
 - appropriateness depends on applications

- Future research:
 - general stochastic intervention on mediators
 - potentially useful for high-dimensional mediators