Discussion of "Decomposition and Interpretation of Treatment Effects in Settings with Delayed Outcomes"

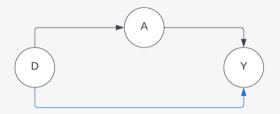
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- RCT or selection-on-observables setting with binary treatment $D \in \{0, 1\}$.
- In practice, often exists a delay between treatment and outcome $Y \in \mathbb{R}$ realization.
- Consequently, study participants may take actions A which can affect Y.

Summary

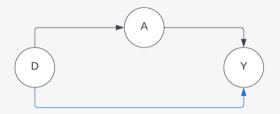
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- Ubiquitous in the literature. Issue: estimands not currently clearly interpretable.
- The paper contributes decomposition of common estimands in published work into:
 - 1. Ceteris paribus effects of D on Y;
 - 2. Indirect effects of A on Y;
 - 3. Selection terms.

- Framework clarifies assumptions needed to interpret estimands as convex combinations of:
 - "controlled" direct effects $E[Y(1, a) Y(0, a)|\Omega];$
 - "controlled" indirect effects $E[Y(0, a) Y(0, 0)|\Omega]$.

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- Conclusion: Required assumptions are potentially untenable in most settings.
- Frequently used estimands typically do not have desired interpretation and may not satisfy strong sign preservation (Simpson's paradox).
- When estimands satisfy conditions above, any combination of "controlled" direct effects is identified via a fully saturated regression.

- Authors examine frequently used specifications and clarify interpretations of estimands.
- Paper is narrowly focused, but also has clear broader implications for mediation analysis.
- Can be viewed as a negative paper, but contains constructive points designs where (A, D) may be allocated randomly or selection-on-observable settings.

- Primary goal is identification of direct effects. Why do they matter?
- If goal is to identify ATE with implementation in mind, seems that the estimand of short regression is the one of interest:

$$\Delta_{short} = E[Y(1) - Y(0)] = E[Y(1, A(1)) - Y(0, A(0))].$$
(1)

- If direct effects are of interest, it appears that careful experimental design may identify direct effects with fewer assumptions in RCTs.

- In particular, if (natural) direct effects $\Delta_{NDE} = E[Y(1, A(1)) Y(0, A(1))]$ is of interest, it may be beneficial to discuss blinding.
- Under appropriate blinding A = A(1), so $\Delta_{NDE} = \Delta_{short}$.
- Examples: 1) Moderna study (placebo vaccines); 2) Fertilizer experiment in Mali (placebo fertilizer).
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- Misclassification error of Y should not affect sign preservation properties (cf. Moderna study).