Supplementary Methods File

Christian A. Maino Vieytes, May A. Baydoun

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1 Four-Way Decomposition Models

1.1 Background

Let Y denote the response, A the exposure (in this case poor oral health problems (POHP; binary–exposed/unexposed)), a its realized value (in the binary exposure case it is "1" or "0"), M the mediator (individual plasma proteomic marker levels or principal component scores), m its realized value, and c the vector of confounders. Under the assumption of no confounding, namely:

i.
$$Y_{am}\coprod A|C$$
 ii. $Y_{am}\coprod M|(A,C)$ iii. $M_a\coprod A|C$ iv. $Y_{am}\coprod A*|C$

that the effect of A on Y is unconfounded conditional on C (i), that the effect of M on Y is unconfounded conditional on (A, C), (ii), the effect of A on M is unconfounded conditional on C (iii), and that any mediator outcome confounders are not affected by the exposure (iv), we can partition the sources of total effect of the model into four components (Equation 1).

$$TE = CDE + INT_{ref} + INT_{med} + PIE \tag{1}$$

We interpret this model to reflect that the total effect of the exposure, A, on the outcome, Y, is a sum of the controlled direct effect (CDE-i.e., the effect of A on Y not due to any interaction or mediation), the reference interaction (INT_{ref} -i.e., the effect of interaction only), the mediated interaction (INT_{med} -i.e., the

effect of interaction and mediation), and the purely indirect effect (PIE–i.e., the indirect effect only). Additional details on the four-way decomposition model are provided in an original publication that we refer the readers to.¹

1.2 Implementation

The process of estimating the components of the partitioned model in Equation 1 requires the fitting of two regression models and then using the parameter estimates for the final computation of the components. A robust summary of the med4way command in Stata is described elsewhere and we refer the reviewers to that commentary though we provide a succinct summary of the implementation and estimation of the effects. Under the assumption of no unmeasured confounding (as detailed above in 1.1) we can estimate, on average, the four components of the model on a population but not the individual-level effects. The two regression models required are provided and include a model for the expectation Y conditioned on the exposure, mediator, and confounders (Equation 2) and a model for the expectation of M conditioned on the exposure and confounders (Equation 3):

$$E[Y|(a, m, c)] = \theta_0 + \theta_1 a + \theta_2 m + \theta_3 a * m + \theta_c c$$
(2)

$$E[M|(a, \mathbf{c})] = \beta_0 + \beta_1 a + \boldsymbol{\beta}_c^T \mathbf{c}$$
(3)

Estimates of these parameters therefore facilitate the direct computation of estimates of the four component sources of variation for the total effect (TE):

$$E[CDE|c] = \theta_1(a - a')$$

$$E[INT_{ref}|c] = \theta_3(\beta_0 + \beta_1 a' + \boldsymbol{\beta}_c^T \boldsymbol{c})(a - a')$$

$$E[INT_{med}|c] = \theta_3\beta_1(a - a')(a - a')$$

$$E[PIE|c] = (\theta_2\beta_1 + \theta_3\beta_1 a')(a - a')$$

where a=1 and a'=0 if the exposure is binary (as it is in this case).³ The models we describe are generalizable and Med4way can handle outcome variables from several distributions (e.g., binomial, log-binomial, Poisson, negative binomial, Weibull, Cox, etc.).² In our analysis, $\mathrm{E}[Y|(a,m,c)]$ and $\mathrm{E}[M|(a,c)]$ are specified as follows:

$$\lambda(t|x, v, \mathbf{z}) = \lambda_0(t) + \theta_1 x + \theta_2 v + \theta_3 x * v + \boldsymbol{\theta}_z^T \mathbf{z}$$
(4)

$$E[V|(x, z)] = \beta_0 + \beta_1 x + \boldsymbol{\beta}_z^T z$$
 (5)

where, in Equation 4, we model the log hazard at time t as a function of x, POHP, v, an individual proteomic biomarker or a principal component score, and z, the vector of confounders/covariates discussed in the manuscript. In Equation 5, we model the expectation of the mediator, V, as a function of POHP and the other covariates using ordinary least squares.

References

- 1. VanderWeele, T. J. A unification of mediation and interaction: a four-way decomposition. *Epidemiology (Cambridge, Mass.)* **25,** 749 (2014).
- 2. Discacciati, A., Bellavia, A., Lee, J. J., Mazumdar, M. & Valeri, L. Med4way: a Stata command to investigate mediating and interactive mechanisms using the four-way effect decomposition 2019.
- 3. VanderWeele, T. J. & Tchetgen Tchetgen, E. J. Mediation analysis with time varying exposures and mediators. *Journal of the Royal Statistical Society Series B: Statistical Methodology* **79**, 917–938 (2017).