



MENDELIAN RANDOMIZATION AND MR-DoC

INTERNATIONAL STATISTICAL GENETICS WORKSHOP

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The University of Melbourne &
Center of Data and Knowledge Integration for Health (CIDACS)

Recap, Mendelian randomization assumptions

Causal inference using twin-design

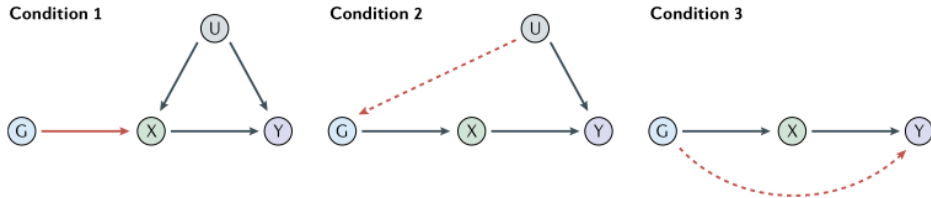
Practical

Summary and take home message

RECAP, MENDELIAN RANDOMIZATION ASSUMPTIONS



- Uses genetic variants as instrumental variables¹
- Helps understand causation, but has strong assumptions²
 1. G (instrument) is robustly associated with X (“relevance”);
 2. G does not share common causes (C) with Y (Outcome) (“independence” or “exchangeability”); and
 3. G affects Y exclusively through its effect on X (“exclusion restriction”).



¹Richmond et al., (2022), “Mendelian Randomization,” *Cold spring harb perspect med*.

²Sanderson et al., (2022), “Mendelian Randomization — Nature Reviews Methods Primers,” *Nature reviews methods primers*.

CAUSAL INFERENCE USING TWIN-DESIGN

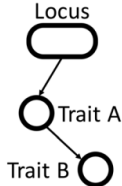


- Genetic variant influences more than one trait
- Horizontal vs Vertical pleiotropy
 - It has a central role in the genetic architecture^a
 - Pleiotropy in over 48% of significant MR,^b with large distortions on MR estimates.

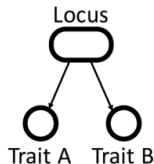
^aJordan et al., (2019), "HOPS," *Genome biology*.

^bVerbanck et al., (2018), "Detection of Widespread Horizontal Pleiotropy in Causal Relationships Inferred from Mendelian Randomization between Complex Traits and Diseases," *Nat genet*.

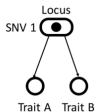
Vertical Pleiotropy



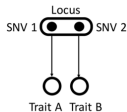
Horizontal Pleiotropy



Biological (single-locus) Horizontal Pleiotropy



LD-Induced Horizontal Pleiotropy



Polygenicity-Induced Horizontal Pleiotropy Causal Loci

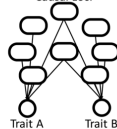


Figure 1: (LD) and polygenicity are expected to contribute to horizontal pleiotropy



- Access to twin data
- Triangulation of findings, confirmation of a causal relationship
- Interest in the variance components or in the background confounding elements



- SEM solutions have recovered exact estimates as 2sls, with caveats^a
 - less convergence in weak instruments
 - slightly worse performance in ML-SEM

^aMaydeu-Olivares et al., (2019), "Instrumental Variables Two-Stage Least Squares (2SLS) Vs. Maximum Likelihood Structural Equation Modeling of Causal Effects in Linear Regression Models," *Structural equation modeling: A multidisciplinary journal*.

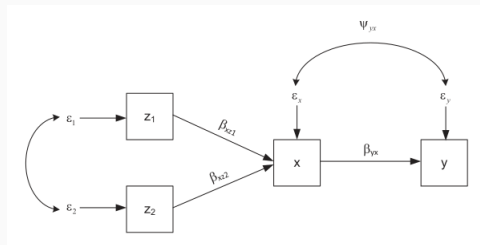


Figure 2: Instrumental Variables Regression (IVR) model that enables drawing causal inferences on the target regression mode



Model specification

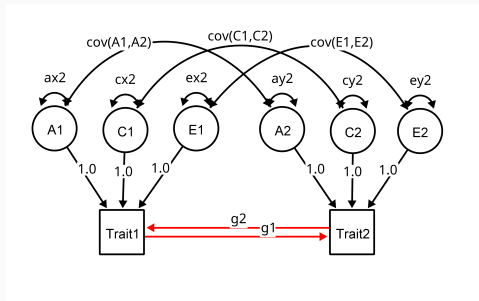


Figure 3: Path diagram representing a Bidirectional DoC for one twin.

$\text{cov}(A1, A2) = r_a * \sqrt{a_{x2} * a_{y2}}$; $\text{cov}(C1, C2) = r_c * \sqrt{c_{x2} * c_{y2}}$; $\text{cov}(E1, E2) = r_e * \sqrt{e_{x2} * e_{y2}}$

- Causal paths are estimated including information from the cross-twin cross-trait correlations.
- Cross-twin covariance between additive genetic effects is 0.5 (not shown) for DZ twins, as DZs are expected to share 50% of the genetic effects.
- Standard SEM symbology is used.

³Neale et al., *Methodology for Genetic Studies of Twins and Families* (New York, NY, US 1992).

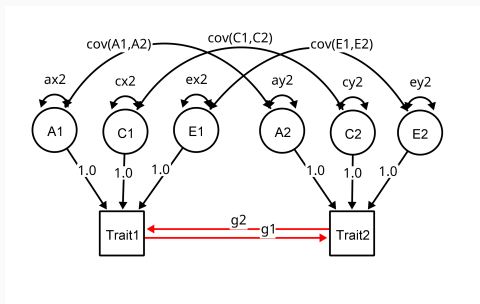


Figure 4: Path diagram representing a Bidirectional DoC for one twin.
 $cov(A_1, A_2) = r_a * \sqrt{a_{x2} * a_{y2}}$; $cov(C_1, C_2) = r_c * \sqrt{c_{x2} * c_{y2}}$; $cov(E_1, E_2) = r_e * \sqrt{e_{x2} * e_{y2}}$

- Bias at the phenotypic level^a
- Bias due to lack of unmodelled E confounding^b
- Better detection of causal paths with different variance component proportions for each phenotype^c

^aGillespie et al., (2003), "Direction of Causation Modeling Between Cross-Sectional Measures of Parenting and Psychological Distress in Female Twins," *Behav genet.*

^bRasmussen et al., (2019), "A Major Limitation of the Direction of Causation Model," *Twin res hum genet.*

^cMaes et al., (2021), "Using Multimodel Inference/Model Averaging to Model Causes of Covariation Between Variables in Twins," *Behav genet.*

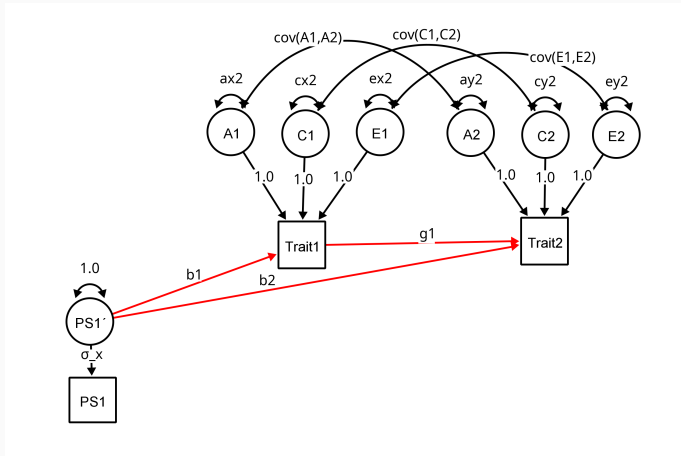


Figure 5: Path diagram for one twin. $\text{cov}(A1,A2) = r_a * \sqrt{a_{x2} * a_{y2}}$; $\text{cov}(C1,C2) = r_c * \sqrt{c_{x2} * c_{y2}}$; $\text{cov}(E1,E2) = r_e * \sqrt{e_{x2} * e_{y2}}$

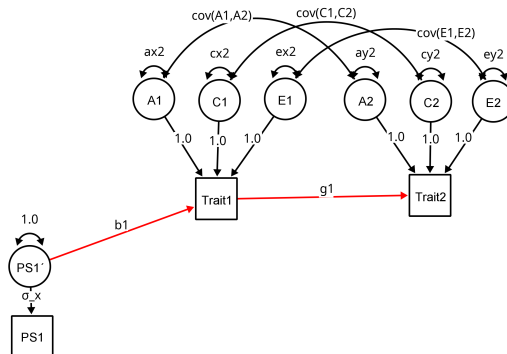
⁴Minică et al., (2018), "Extending Causality Tests with Genetic Instruments," *Behav genet.*

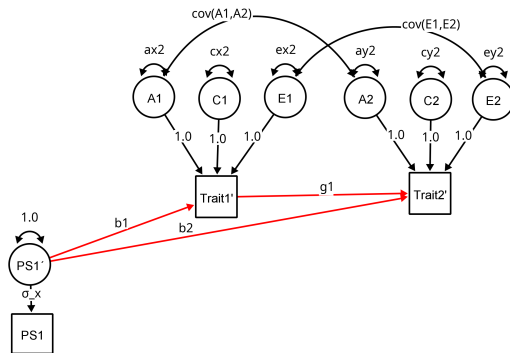


x	aX	cX	eX	aY	cY	eY	ra	rc	re	b1	b2	g1	Id
fr	fr	fr	fr	fr	fr	fr	fr	fr	fr	fr	fr	fr	No
fr	fr	fr	fr	fr	fr	fr	fr	fr	0	fr	fr	fr	Yes
fr	fr	fr	fr	fr	fr	fr	fr	fr	fr	fr	0	fr	Yes
fr	fr	fr	fr	fr	fr	fr	fr	0	fr	fr	fr	fr	Yes
fr	fr	0	fr	fr	0	fr	fr	0	fr	fr	fr	fr	No
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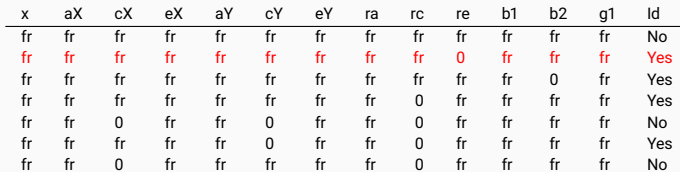


x	aX	cX	eX	aY	cY	eY	ra	rc	re	b1	b2	g1	Id
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fr	fr	fr	fr	fr	fr	fr	fr	fr	0	fr	fr	fr	Yes
fr	fr	fr	fr	fr	fr	fr	fr	fr	fr	fr	0	fr	Yes
fr	fr	fr	fr	fr	fr	fr	fr	0	fr	fr	fr	fr	Yes
fr	fr	0	fr	fr	0	fr	fr	0	fr	fr	fr	fr	No
fr	fr	fr	fr	fr	0	fr	fr	0	fr	fr	fr	fr	Yes
fr	fr	0	fr	fr	fr	fr	fr	0	fr	fr	fr	fr	No



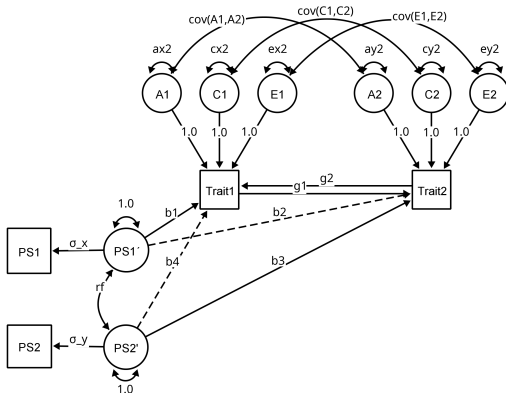






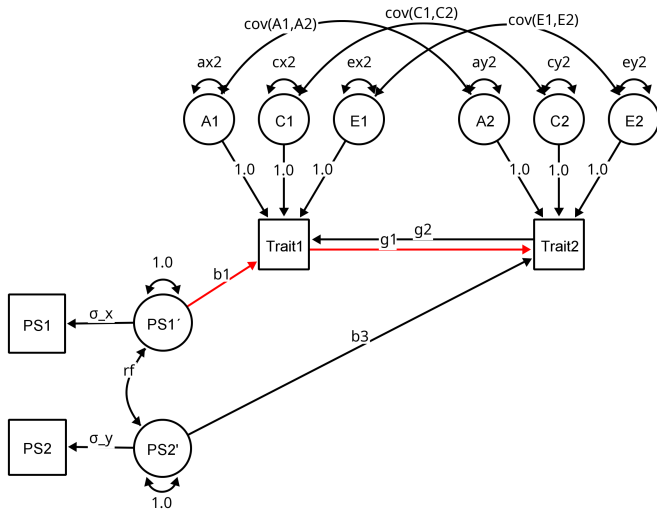


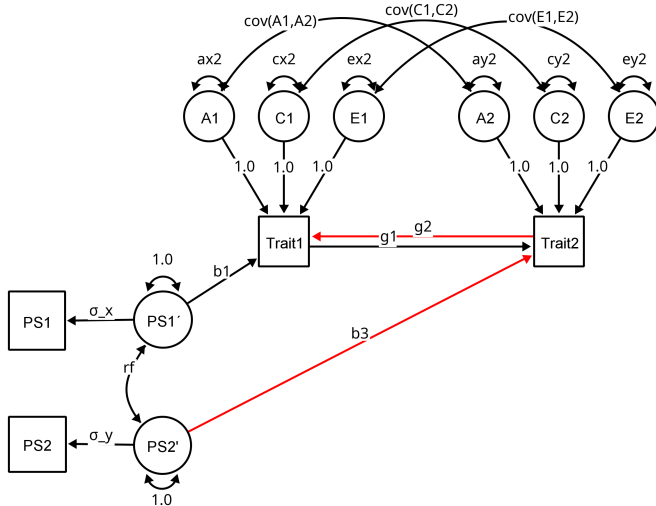
Model specification



- Path diagram of the MR-DoC2 model for an individual.^a
- The model includes the effects of additive genetic (A), common environment (C) and unique environment (E) factors for both X and Y, and their effects may correlate.

^aCastro-de-Araujo et al., (2023), "MR-DoC2," *Behav genet.*







- MR-DoC g_1 is biased if modeled $re = 0$ (or $cov(E1, E2) = 0$), when re is **not equal** 0 in the data.
- MR-DoC2 requires very large sample sizes to detect g_1, g_2
- MR-DoC2 g_1, g_2 unbiased in the presence of measurement error

⁵Castro-de-Araujo et al., "Power, Measurement Error, and Pleiotropy Robustness in Twin-Design Extensions to Mendelian Randomization," Available at <https://www.researchsquare.com> (accessed November 6, 2023).

PRACTICAL



Make sure you are in your home folder by typing:

`pwd`

Create a directory to hold today's work by typing:

`mkdir mrdoc`

Change your directory to the new folder:

`cd mrdoc`

Copy files, don't forget the dot at the end

`cp /faculty/luis/2024/mrdoc/* .`



Beta matrix

```
BE <- mxMatrix(name = "BE", type = "Full", nrow=3, ncol = 3, byrow = TRUE,
  labels = c(NA, "g2", "b1",
             "g1", NA, "b2",
             NA, NA, NA),
  free = c(FALSE, FALSE, TRUE,
           TRUE, FALSE, TRUE,
           FALSE, FALSE, FALSE),
  dimnames = list(c("X", "Y", "iX"),
                  c("X", "Y", "iX")))
```



A, C, E variances

```
# ACE decomposition
A <- mxMatrix(name = 'A', type='Symm', nrow=3,ncol = 3,byrow = TRUE,
              labels=c("ax2",  "covA", NA,
                        "covA", "ay2",  NA,
                        NA,      NA,    "sigma_x"),
              free=c(TRUE, TRUE, FALSE,
                     TRUE, TRUE, FALSE,
                     FALSE,FALSE,TRUE),
              dimnames = list(c("X", "Y", "iX"),
                              c("X", "Y", "iX")))
```



A filter matrix is required to drop PRSs from twin 2 in MZs

```
# The filter matrix is used to remove the PRS from MZs,  
# if we kept it the matrix would become redundant resulting  
# in Hessian not positive  
filter <- mxMatrix(name = 'filter', type='Full', nrow=5, ncol=6, free=FALSE,  
  byrow = TRUE,  
  values=c(1,0,0,0,0,0,  
           0,1,0,0,0,0,  
           0,0,1,0,0,0,  
           0,0,0,1,0,0,  
           0,0,0,0,1,0))
```



- In the script you will find the pipe operator from R base `|>`, this is done to achieve concise blocks of code while still being expressive.
- The idea is that the resulting object from one instruction is passed on to the next instruction. Example:

```
unrel4 <- unrel3 |>  
  omxSetParameters(name = "no_causal", label="g1", free = FALSE,  
                    values = 0) |>  
  mxRun()
```

- In the case above, `omxSetParameters()` returns a model with modified parameters and this model is then passed to `mxRun()`, which returns the same model after estimation to the object `unrel4` using the assignment operator `<-`.



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Aims:

1. How to setup a simulation in OpenMx using `mxGenerateData`
2. How to specify a two-stage least squares test in OpenMx
3. Check that MR-DoC recovers the same estimates from 2sls test
4. Check that in the presence of horizontal pleiotropy estimates are biased in the 2sls test

The script will exemplify four scenarios:

- no pleiotropy between the instrument and the outcome (Scenario 1),
- where there is pleiotropy (Scenario 2),
- presence of a bidirectional causal effect (Scenario 3),
- absence of a causal effect (Scenario 4)



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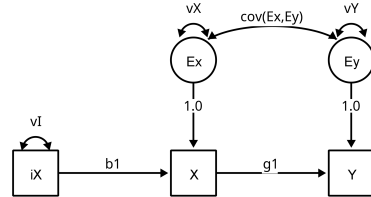
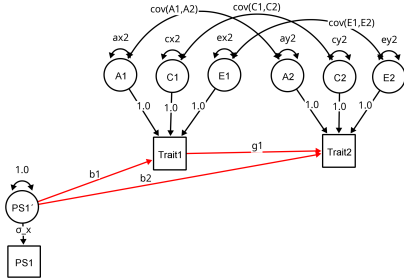
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SUMMARY AND TAKE HOME MESSAGE



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