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# Half-Trek Criterion for Identifiability of Latent Variable Models Rina Foygel Barber<sup>1</sup>, Mathias Drton<sup>2</sup>, Nils Sturma<sup>2</sup>, Luca Weihs<sup>3</sup>

#### 1. Motivation

**Assumption:** Known causal structure between observed and latent variables.



#### 2. Setup

Linear structural equation model with observed variables  $X = (X_v)_{v \in V}$  and latent variables  $L = (L_h)_{h \in \mathcal{L}}$ :

 $X = \Lambda^{\top} X + \Gamma^{\top} L + \varepsilon$ 

### 3. Rational Identifiability

**Given:** Latent-factor graph  $G = (V \dot{\cup} \mathcal{L}, D)$ .

Every latent-factor graph G yields a parametrization of the observed covariance matrix:

Latent-factor graph with 5 observed nodes and one latent node h.

**Aim:** Identify the direct causal effects between the observed variables based on the observed covariance matrix. (identify = uniquely recover)

#### Main contributions:

- Sufficient condition for rational identifiability in a linear setting.
- Applicable in settings where latent variables may also have dense effects on many or even all of the observables.
- Recursive polynomial time algorithm. (when bounding a matrix rank in a search step)

- Sparsity: Parameter matrices  $\Lambda$  and  $\Gamma$  are supported over the edge set D of a directed graph  $G = (V \dot{\cup} \mathcal{L}, D)$ .
- Latent-factor assumption: All nodes in  $\mathcal{L}$  are source nodes of G.
- Independence of the latent factors and the error terms:  $Var[\varepsilon] =: \Omega_{diag}$  is diagonal and  $\operatorname{Var}[L] = I.$

Latent covariance matrix:

 $\Omega \equiv \operatorname{Var}[\Gamma^{\top}L + \varepsilon]$  $= \operatorname{Var}[\varepsilon] + \Gamma^{\top} \operatorname{Var}[L] \Gamma = \Omega_{\operatorname{diag}} + \Gamma^{\top} \Gamma.$ 

Observed covariance matrix:

 $\Sigma \equiv \operatorname{Var}[X] = (I - \Lambda)^{-\top} \Omega (I - \Lambda)^{-1}.$ 

 $\varphi_G : (\Lambda, \Gamma, \Omega_{\mathsf{diag}}) \longmapsto \Sigma \equiv \mathsf{Var}[X].$ 

**Definition.** The model given by G is rationally identifiable if there is a rational map  $\psi_G$  such that

 $\psi_G \circ \varphi_G(\Lambda, \Gamma, \Omega_{\mathsf{diag}}) = \Lambda$ 

for 'almost all'  $(\Lambda, \Gamma, \Omega_{diag})$ .

Remark. Always solvable via Gröbner basis computations. • Double-exponential complexity.

• Only feasible on small graphs.

# Software

#### SEMID

An R-package for parameter identifiability in linear structural equation models.

 $\rightarrow$  available on CRAN  $\bigcirc$  and GitHub  $\bigcirc$ 

# 4. Key Idea

Use algebraic relations in latent covariance

# 5. LF Half-Trek Criterion

**Definition.** A half-trek from node v to node w is

# **Example I**

matrix.

Observe that

 $\Sigma = (I - \Lambda)^{-\top} \Omega (I - \Lambda)^{-1}$  $\iff \Omega = (I - \Lambda)^T \Sigma (I - \Lambda).$ 

Algebraic relations among entries of  $\Omega = \Omega_{diag} + \Omega_{diag}$  $\Gamma^{+}\Gamma$  yield relations among entries of  $\Lambda$  and  $\Sigma$ :

 $f(\Omega) = 0 \iff f((I - \Lambda)^T \Sigma (I - \Lambda)) = 0.$ 

#### Observation:

The latent covariance matrix may be sparse and feature low-rank structure:

 $\Omega = \Omega_{\mathsf{diag}} + \Gamma^{\top}\Gamma = \Omega_{\mathsf{diag}} + \sum_{h \in \mathcal{L}} \gamma_h \gamma_h^{\top}$ = diag + sum of sparse rank 1 matrices.

 $\longrightarrow$  We exploit algebraic relations that are vanishing off-diagonal sub-determinants of  $\Omega$ .

#### Example:



#### a path of the form

$$v \rightarrow x_1 \rightarrow w$$
 or  $v \rightarrow x_1 \rightarrow x_n$ 





**Definition.** Let  $v \in V$  and  $Y, Z \subseteq V \setminus \{v\}$ and  $H \subseteq \mathcal{L}$ . The triple (Y, Z, H) satisfies the late<u>nt-factor half-trek criterion</u> (LF-HTC) for v if

1. |Y| = |pa(v)| + |Z| and |Z| = |H|,

 $\mathbf{2.} Y \cap (Z \cup \{v\}) = \emptyset,$ 

**3.**  $[\operatorname{pa}(Y) \cap \operatorname{pa}(Z \cup \{v\}) \cap \mathcal{L}] \subseteq H$ ,

4. There is system of half-treks from Y to  $pa(v) \cup v$ Z without sided intersection and all half-treks ending in Z have form  $y \leftarrow h \rightarrow z$  for  $h \in H$ .

**Theorem.** If the triple (Y, Z, H) satisfies the LF-HTC for  $v \in V$ , then column  $\Lambda_{*,v}$  is a rational function of the observed covariance matrix  $\Sigma$ ,



 $v \in \{1, 4\}$ : Trivially,  $\Lambda_{*,1} = \Lambda_{*,4} = 0$ .

<u>v = 3</u>: Take  $Y = \{1, 2\}, Z = \{4\}, H = \{h\}$ . (ii)  $Y \cap (Z \cup \{3\}) = \{1, 2\} \cap \{3, 4\} = \emptyset$ , (iv)  $1 \leftarrow h \rightarrow 4, 2 \equiv 2$ 

v = 2 and v = 5: Can find (Y, Z, H) similarly.

 $\implies$  The model is LF-HTC-identifiable, that is, the parameters  $\lambda_{12}$ ,  $\lambda_{23}$ , and  $\lambda_{45}$  are recovered by rational functions in the entries of  $\Sigma$ .

# **Example II**





(a) Rationally identifiable.

(b) Generically finite-

We have the following relations among  $\Lambda$  and  $\Sigma$ :

 $\det\left(\left[(I-\Lambda)^T\Sigma(I-\Lambda)\right]_{\{1,2\},\{3,4\}}\right)$  $= \lambda_{23}\sigma_{12}\sigma_{24} - \lambda_{23}\sigma_{14}\sigma_{22} - \sigma_{13}\sigma_{24} + \sigma_{14}\sigma_{23} = 0,$ 

which we can then solve for  $\lambda_{23}$ .

the columns  $(\Lambda_{*,z})_{z\in Z}$  and the columns  $\Lambda_{*,y}$  for those  $y \in Y$  that can be reached from  $Z \cup \{v\}$ using a half-trek that avoids H.

**Algorithm.** Recursively cycle through nodes vand search for LF-HTC triples that allow solving for  $\Lambda_{*,v}$ . Network-flow setup finds LF-HTC triples in polynomial time under a bound on |Z| = |H|.

to-one but not rationally identifiable.



(c) Generically infinite-toone.

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