

# Identifiability of linear structural equation models with homoscedastic errors using algebraic matroids

Jun Wu<sup>1</sup>, Mathias Drton<sup>1</sup>, Benjamin Hollering<sup>2</sup>

<sup>1</sup>Technical University of Munich, Germany;  
TUM School of Computation, Information and Technology,  
Department of Mathematics and Munich Data Science Institute

<sup>2</sup>Research Group in Nonlinear Algebra,  
Max Planck Institute for Mathematics in the Sciences

German Probability and Statistics Day, March 2023

- 1 Linear Structural Equation Models
- 2 Structural Identifiability
- 3 Jacobian Matroid
- 4 Outdegree Proposition and etc.
- 5 Identifiability Results
- 6 Computational Checks for  $|V| \leq 6$

- 1 Linear Structural Equation Models
- 2 Structural Identifiability
- 3 Jacobian Matroid
- 4 Outdegree Proposition and etc.
- 5 Identifiability Results
- 6 Computational Checks for  $|V| \leq 6$

# Linear Structural Equation Models

- Random vector  $X = (X_i : i \in V)$  solves

$$X = \Lambda^T X + \varepsilon, \quad \text{Var}[\varepsilon] = \Omega.$$

- We consider **homoscedastic errors**,  $\Omega = \omega \cdot I$ , and then focus on the precision matrix:

$$\psi_G(\Lambda, s) = \Sigma^{-1} = s(I - \Lambda)(I - \Lambda)^T, \quad s = \frac{1}{\omega}.$$

- The **linear homoscedastic Gaussian model** given by a directed graph  $G = (V, D)$  is

$$M_G = \left\{ s(I - \Lambda)(I - \Lambda)^T : \Lambda \in \mathbb{R}_{\text{reg}}^D, s > 0 \right\},$$

where  $\mathbb{R}_{\text{reg}}^D = \left\{ \Lambda \in \mathbb{R}^{V \times V} : \Lambda_{ij} = 0 \text{ if } i \rightarrow j \notin D, I - \Lambda \text{ invertible} \right\}$ .

# Linear Structural Equation Models: Example

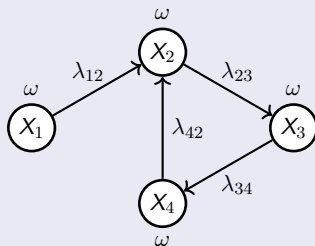
## Example 1

$$X_1 = \varepsilon_1$$

$$X_2 = \lambda_{12}X_1 + \lambda_{42}X_4 + \varepsilon_2$$

$$X_3 = \lambda_{23}X_2 + \varepsilon_3$$

$$X_4 = \lambda_{34}X_3 + \varepsilon_4$$



$$\Lambda = \begin{pmatrix} 0 & \lambda_{12} & 0 & 0 \\ 0 & 0 & \lambda_{23} & 0 \\ 0 & 0 & 0 & \lambda_{34} \\ 0 & \lambda_{42} & 0 & 0 \end{pmatrix}, \quad s = \frac{1}{\omega}$$

The graph is **simple** and the SEM is **non-recursive** ( $\exists$  cycle)

- 1 Linear Structural Equation Models
- 2 Structural Identifiability**
- 3 Jacobian Matroid
- 4 Outdegree Proposition and etc.
- 5 Identifiability Results
- 6 Computational Checks for  $|V| \leq 6$

# Structural Identifiability

Question: Can two different graphs have the same model?

- Basic case: **Markov equivalent classes** of DAGs (directed acyclic graphs)
- Homoscedastic errors: within the class of **DAGs**, the graph  $G$  is known to be identifiable  
[Chen, Drton, and Wang 2019; Peters and Bühlmann 2014]
- Identifiability results when cycles allowed?

## Definition

Let  $\{M_i\}_{i=1}^k$  be a finite set of algebraic statistic models given by subsets of  $\mathbb{R}^m$ . The indices  $i$ 's are **generically identifiable** if for each pair of  $(i_1, i_2)$ ,

$$\dim(M_{i_1} \cap M_{i_2}) < \max(\dim(M_{i_1}), \dim(M_{i_2})).$$

# Structural Identifiability

How to compare two models?

- Traditional method: Groebner basis (equi.)
- We use: Jacobian matroid (suff., related to graphical criteria)

## Our contributions

- Derive graphical criteria certifying two simple directed graphs have distinguishable models
- Give subclasses of graphs that are generically identifiable
- Computational checks for small-size graphs

For a simple directed graph  $G = (V, D)$ ,

$$\dim(M_G) := \text{rank}(J(\psi_G)) = |D| + 1.$$



- 1 Linear Structural Equation Models
- 2 Structural Identifiability
- 3 Jacobian Matroid**
- 4 Outdegree Proposition and etc.
- 5 Identifiability Results
- 6 Computational Checks for  $|V| \leq 6$

# Jacobian: Example

## Example 2

$G = (V, D)$ , with  $V = \{1, 2, 3, 4\}$  and  $D = \{(1, 2), (2, 4), (1, 3), (3, 4)\}$

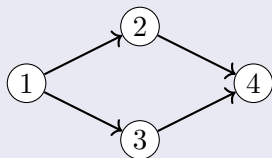


Figure: Example 2

$J(\psi_G) :$

$$\begin{pmatrix} K_{11} & K_{22} & K_{33} & K_{44} & K_{12} & K_{23} & K_{34} & K_{13} & K_{24} & K_{14} \\ 2s\lambda_{12} & 0 & 0 & 0 & -s & 0 & 0 & 0 & 0 & 0 \\ 2s\lambda_{13} & 0 & 0 & 0 & 0 & 0 & 0 & -s & 0 & 0 \\ 0 & 2s\lambda_{24} & 0 & 0 & 0 & s\lambda_{34} & 0 & 0 & -s & 0 \\ 0 & 0 & 2s\lambda_{34} & 0 & 0 & s\lambda_{24} & -s & 0 & 0 & 0 \\ 1 + \lambda_{12}^2 + \lambda_{13}^2 & 1 + \lambda_{24}^2 & 1 + \lambda_{34}^2 & 1 & -\lambda_{12} & \lambda_{24}\lambda_{34} & -\lambda_{34} & -\lambda_{13} & -\lambda_{24} & 0 \end{pmatrix} \begin{matrix} \lambda_{12} \\ \lambda_{13} \\ \lambda_{24} \\ \lambda_{34} \\ s \end{matrix}$$

$$\text{rank}(J_{\{44,12,34,13,24\}}) = 5$$

## Definition

Suppose  $M = \text{Im}(\phi)$  with parametrization  $\phi(\theta) = (\phi_1(\theta), \dots, \phi_r(\theta))$ . Let

$$J(\phi) = \left( \frac{\partial \phi_j}{\partial \theta_i} \right), 1 \leq i \leq d, 1 \leq j \leq r$$

be the Jacobian of  $\phi$ . Then the **Jacobian matroid** of model  $M$  is the matroid  $\mathcal{M}(\phi) = (E, \mathcal{I})$ , where

- $E = [r]$ , the set of column indices
- A set  $S \in \mathcal{I} \subseteq 2^E$  is called an independent set
- The columns of  $J(\phi)$  indexed by  $S$  are linearly independent over the fraction field  $\mathbb{R}(\theta)$

## Example 3

$$\phi(t_1, t_2, t_3) = (t_1, -t_1^2, t_1 t_2 + t_3^2),$$

$$J = \begin{bmatrix} 1 & -2t_1 & t_2 \\ 0 & 0 & t_1 \\ 0 & 0 & 2t_3 \end{bmatrix}.$$

- $E = \{1, 2, 3\}$
- The independent sets are

$$\emptyset, \{1\}, \{2\}, \{3\}, \{1, 3\}, \{2, 3\}$$

## Proposition [Hollering and Sullivant 2021]

Let  $M_1$  and  $M_2$  be two parameterized models in  $\mathbb{R}^m$  with parameterization  $\psi_1$  and  $\psi_2$ . Assuming without loss of generality that  $\dim(M_1) \geq \dim(M_2)$ , if there exists a subset  $S$  of the columns such that

$$S \in \mathcal{M}(\psi_2) \setminus \mathcal{M}(\psi_1),$$

then  $\dim(M_1 \cap M_2) < \min(\dim(M_1), \dim(M_2))$ .

- A **sufficient condition** for generic identifiability
- $M_1, M_2$  exchangeable when  $\dim(M_1) = \dim(M_2)$

- 1 Linear Structural Equation Models
- 2 Structural Identifiability
- 3 Jacobian Matroid
- 4 Outdegree Proposition and etc.
- 5 Identifiability Results
- 6 Computational Checks for  $|V| \leq 6$

# Outdegree Proposition

## Proposition 1 (Outdegree proposition)

Let  $G_1, G_2$  be non-complete simple directed graphs. If  $\exists i$  s.t.  $|Ch_1(i)| \neq |Ch_2(i)|$  then  $\mathcal{M}(\psi_1) \neq \mathcal{M}(\psi_2)$ .

## Example 4

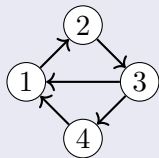


Figure:  $G_1$

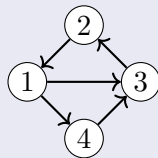


Figure:  $G_2$

- $G_1$  has outdegree sequence  $\{1, 1, 2, 1\}$
- $G_2$  has outdegree sequence  $\{2, 1, 1, 1\}$

# Example: Outdegree Proposition Not Applicable

## Example 5

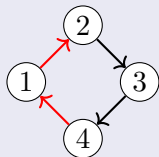


Figure:  $G_1$

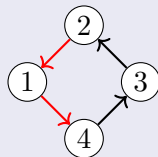


Figure:  $G_2$

Let  $S = \{22, 33, 23, 34, 14\}$ ,

$$J_S^1 = \begin{pmatrix} K_{22} & K_{33} & K_{23} & K_{34} & K_{14} \\ 0 & 0 & 0 & 0 & 0 \\ 2s\lambda_{23} & 0 & -s & 0 & 0 \\ 0 & 2s\lambda_{34} & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & -s \\ 1 + \lambda_{23}^2 & 1 + \lambda_{34}^2 & -\lambda_{23} & -\lambda_{34} & -\lambda_{41} \end{pmatrix} \begin{matrix} \lambda_{12} \\ \lambda_{23} \\ \lambda_{34} \\ \lambda_{41} \\ s \end{matrix}, \quad \text{rank}(J_S^1) = 4,$$

$$J_S^2 = \begin{pmatrix} K_{22} & K_{33} & K_{23} & K_{34} & K_{14} \\ 2s\lambda_{21} & 0 & 0 & 0 & 0 \\ 0 & 2s\lambda_{32} & -s & 0 & 0 \\ 0 & 0 & 0 & -s & 0 \\ 0 & 0 & 0 & 0 & -s \\ 1 + \lambda_{21}^2 & 1 + \lambda_{32}^2 & -\lambda_{32} & -\lambda_{43} & -\lambda_{14} \end{pmatrix} \begin{matrix} \lambda_{21} \\ \lambda_{32} \\ \lambda_{43} \\ \lambda_{14} \\ s \end{matrix}, \quad \text{rank}(J_S^2) = 5.$$



# Parentally Closed set condition

## Definition

A set  $L \in \text{ne}(i)$  is called **parentally closed** w.r.t node  $i$  if  $\text{pa}(L) \cap \text{ne}(i) \subseteq L$ .

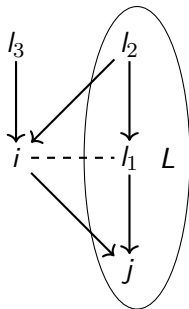


Figure: A parentally closed set

# Parentally Closed Set Condition

## Proposition 2 (Parentally closed set)

Let  $G_1 = (V, D_1)$ ,  $G_2 = (V, D_2)$  be two simple directed graphs, both not complete. For any node  $i$ , there are two collections of parentally closed sets  $\mathcal{L}_i^1, \mathcal{L}_i^2$ , corresponding to  $G_1$  and  $G_2$ . If there is a set  $L \in \mathcal{L}_i^k$  such that  $|Ch_k(i) \cap L| > |Ch_{3-k}(i) \cap L|$ ,  $k \in \{1, 2\}$ , then  $G_1$  and  $G_2$  have different matroids.

## Corollary 2.1 (Transitive triangle-free)

If two different non-complete simple graphs do not contain **transitive triangles** ( $i \rightarrow j \rightarrow k$  and  $i \rightarrow k$ ), then they have different matroids.

(Every parentally closed set in a transitive triangle-free graph is a singleton!)

- 1 Linear Structural Equation Models
- 2 Structural Identifiability
- 3 Jacobian Matroid
- 4 Outdegree Proposition and etc.
- 5 Identifiability Results**
- 6 Computational Checks for  $|V| \leq 6$

# Identifiability Results

## Theorem 1

Let  $\mathcal{G}$  be the collection of **non-complete simple directed graphs**. If the collection satisfies **one** of the following conditions, then the models set of graphs in  $\mathcal{G}$  is generically identifiable.

- (i) Every graph  $G \in \mathcal{G}$  has a **unique outdegree sequence**
- (ii) Every graph  $G \in \mathcal{G}$  does not contain a **transitive triangle** ( $i \rightarrow j \rightarrow k$  and  $i \rightarrow k$ )

## Theorem 2

A DAG and a cyclic graph (generically) generate **different distributions** under homoscedastic errors condition.

## Theorem 3

The collection of all **non-complete simple directed graphs** with **at least 1 source node**, and whose **strongly connected components** contain no transitive triangles are generically identifiable.

- 1 Linear Structural Equation Models
- 2 Structural Identifiability
- 3 Jacobian Matroid
- 4 Outdegree Proposition and etc.
- 5 Identifiability Results
- 6 Computational Checks for  $|V| \leq 6$

# Computational checks for $|V| \leq 6$

## Methods

- $|V| \leq 5$ : Complete symbolic rank checks
- $|V| = 6$ : Brute force check: **extremely time-consuming!**

To relieve the issue:

- Comparisons: within the subclasses indexed by **outdegree sequences**
- Parameters: **random integers**

## Results




- Most of the simple directed graphs have **unique matroids**
- Some graph pairs have the same matroids, but **can be distinguished** by entries in the precision matrix
- Compatible with parental closed set condition checks



- Natural extension of equal variance (homoscedastic) error assumption from DAGs to directed cyclic graphs
- Partial identifiability results of linear homoscedastic Gaussian models
- Some side-results in algebra



- Results are not strong enough to cover all cases
- Computation capacity is currently limited to 6-node graphs

-  Chen, Wenyu, Mathias Drton, and Y Samuel Wang (Sept. 2019). “On causal discovery with an equal-variance assumption”. In: *Biometrika* 106.4, pp. 973–980.
-  Hollering, Benjamin and Seth Sullivant (2021). “Identifiability in phylogenetics using algebraic matroids”. In: *Journal of Symbolic Computation* 104, pp. 142–158.
-  Peters, Jonas and Peter Bühlmann (2014). “Identifiability of Gaussian structural equation models with equal error variances”. In: *Biometrika* 101.1, pp. 219–228.