

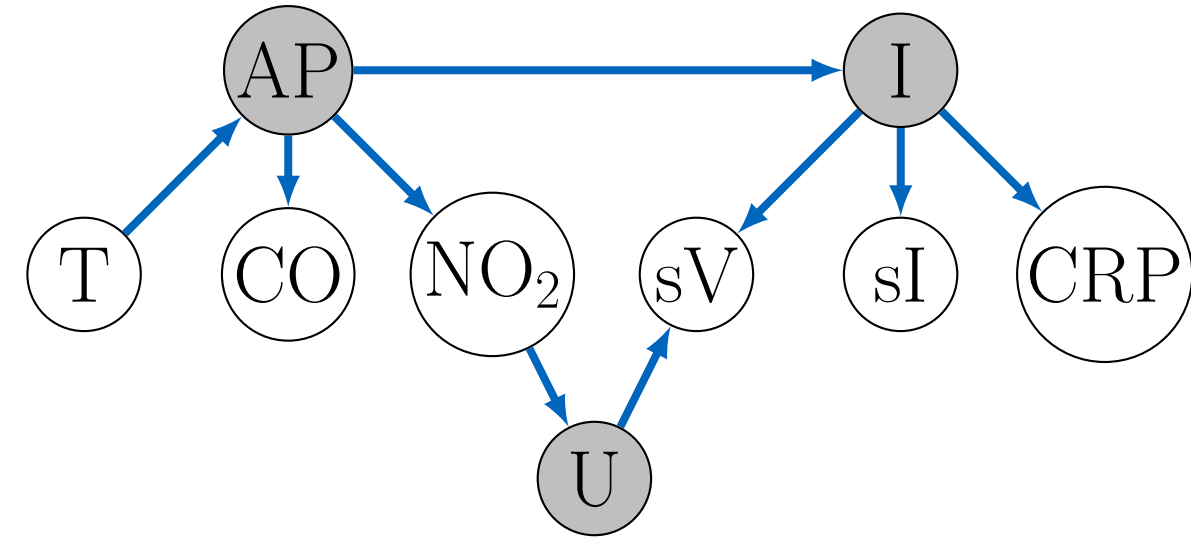
TREK-BASED PARAMETER IDENTIFICATION FOR LINEAR CAUSAL MODELS WITH ARBITRARILY STRUCTURED LATENT VARIABLES

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Motivation & Setup

Directed graph captures causal relations between variables. Example:



Setup:

• Directed graph $G = (V, D)$ with observed and latent nodes $V = \mathcal{O} \sqcup \mathcal{L}$.

• Random vector $X = (X_v)_{v \in V}$ satisfies

$$X_v = \sum_{w \in \text{pa}(v)} \lambda_{vw} X_w + \varepsilon_v,$$

where $\varepsilon = (\varepsilon_v)_{v \in V}$ are jointly independent with $\text{Var}[\varepsilon_v] = \phi_v$.

• Matrix form: $X = \Lambda^\top X + \varepsilon \implies \Sigma := \text{Cov}[X] = (I - \Lambda)^{-\top} \Phi (I - \Lambda)^{-1}$.

Goal: Identify *semi-direct effects* between observed variables from $\text{Cov}[X]$.

Semi-direct effect $v \rightsquigarrow w$ corresponds to sum over all paths of the form

$$v \rightarrow w \quad \text{or} \quad v \rightarrow h_1 \rightarrow \dots \rightarrow h_k \rightarrow w \quad \text{with } h_i \in \mathcal{H} \text{ for all } i.$$

\implies Identify $\bar{\Lambda} := \Lambda_{\mathcal{O}, \mathcal{O}} + \Lambda_{\mathcal{O}, \mathcal{L}} (I - \Lambda_{\mathcal{L}, \mathcal{L}})^{-1} \Lambda_{\mathcal{L}, \mathcal{O}}$.

Main contributions:

- Sufficient graphical condition for rational identifiability of $\bar{\Lambda}$.
- Efficient sound and complete algorithm.
- Allows to transform model to simpler measurement model.

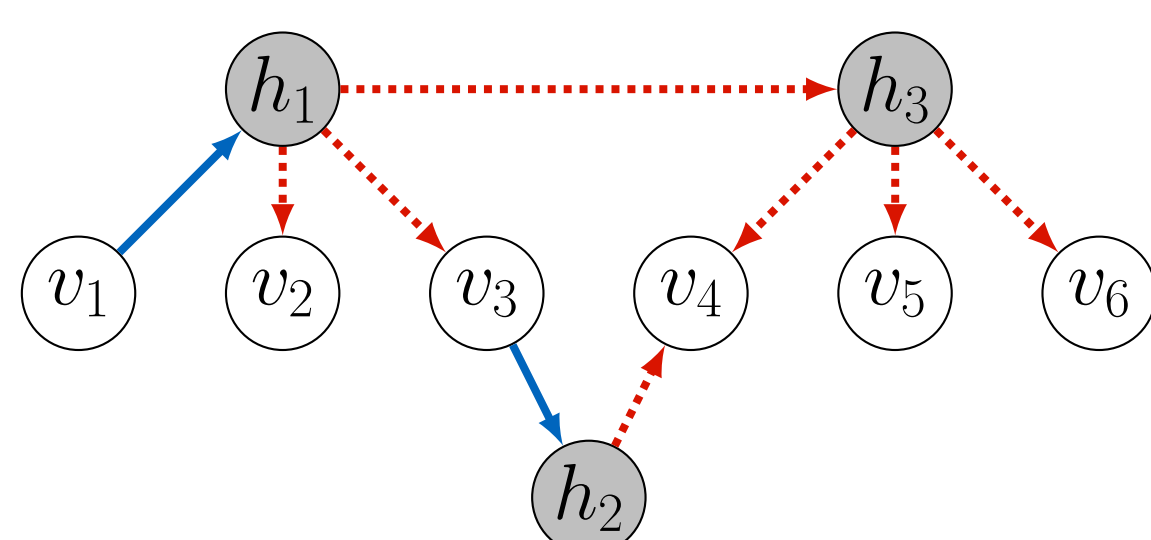
Factorization of the Covariance Matrix

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

Observation:

Ω is covariance matrix in the model given by **subgraph** G_{lat} obtained from removing all edges $v \rightarrow w$ with $v \in \mathcal{O}$.

Example:



Key Idea

Fix $v \in \mathcal{O}$ and consider $\bar{\Lambda}_{\bar{\text{pa}}(v), v}$. Find linear equation system

$$[(I - \bar{\Lambda})^\top \Sigma]_{Y, \bar{\text{pa}}(v)} \quad \Omega_{Y, Z} \cdot \begin{pmatrix} \bar{\Lambda}_{\bar{\text{pa}}(v), v} \\ \psi \end{pmatrix} = [(I - \bar{\Lambda})^\top \Sigma]_{Y, v} \quad (*)$$

such that the matrix on the left-hand side is *invertible* and that $\Omega_{Y, Z}$ and suitable entries in $\bar{\Lambda}$ are *already certified to be identifiable* from earlier calculations.

Requires new proof technique:

Trek rule: $\Sigma_{vw} = [(I - \Lambda)^{-\top} \Phi (I - \Lambda)^{-1}]_{vw}$ is sum over 'treks' $v \leftarrow \dots \leftarrow k \rightarrow \dots \rightarrow w$.

$$\implies \Sigma_{vw} = \sum_{\pi \in \mathcal{T}(v, w)} \phi_k \prod_{x \rightarrow y \in \pi} \lambda_{xy}.$$

Similarly:

• Ω_{vw} is sum over treks in G_{lat} .

• $[(I - \bar{\Lambda})^\top \Sigma]_{vw} = [\Omega (I - \bar{\Lambda})^{-1}]_{vw}$ is sum over treks with left part in G_{lat} .

Trek-Separation in Subgraphs. If there exists a system of treks with no sided intersection from Y to $\bar{\text{pa}}(v) \cup Z$ in G such that the left part of every trek only takes edges in G_{lat} , and the right part of every trek ending in Z only takes edges in G_{lat} , then we have generically

$$\det([(I - \bar{\Lambda})^\top \Sigma]_{Y, \bar{\text{pa}}(v)} \quad \Omega_{Y, Z}) \neq 0.$$

Latent Subgraph Criterion

Definition. Given a node $v \in \mathcal{O}$, the 4-tuple $(Y, Z, H_1, H_2) \in 2^{\mathcal{O} \setminus \{v\}} \times 2^{\mathcal{O} \setminus \{v\}} \times 2^{\mathcal{L}} \times 2^{\mathcal{L}}$ satisfies the *latent-subgraph criterion* (LSC) for v if

- $|Y| = |\bar{\text{pa}}(v)| + |Z|$ and $|Z| = |H_1| + |H_2|$ with $Z \cap \bar{\text{pa}}(v) = \emptyset$,
- $Y \cap (Z \cup \{v\}) = \emptyset$ and (H_1, H_2) trek separates Y and $Z \cup \{v\}$ in the latent subgraph G_{lat} ,
- there exists a system of treks with no sided intersection from Y to $\bar{\text{pa}}(v) \cup Z$ in G such that the left part of every trek only takes edges in G_{lat} , and the right part of every trek ending in Z only takes edges in G_{lat} .

Interpretation:

• (i) - (iii) ensure that there is $\psi \in \mathbb{R}^{|Z|}$ s.t. $\Omega_{Y, Z} \cdot \psi = \Omega_{Y, v}$.

$$\implies [(I - \bar{\Lambda})^\top \Sigma (I - \bar{\Lambda})]_{Y, v} - \Omega_{Y, Z} \cdot \psi = 0.$$

$$\implies (*) \text{ holds.}$$

• (iii) ensures invertibility.

Theorem. If the tuple $(Y, Z, H_1, H_2) \in 2^{\mathcal{O} \setminus \{v\}} \times 2^{\mathcal{O} \setminus \{v\}} \times 2^{\mathcal{L}} \times 2^{\mathcal{L}}$ satisfies the LSC for $v \in \mathcal{O}$, then $\bar{\Lambda}_{*, v}$ is a rational function of the observed covariance matrix Σ , the semi-direct effects $(\bar{\Lambda}_{*, z})_{z \in Z}$, and the semi-direct effects $(\bar{\Lambda}_{*, y})_{y \in Y}$ for those $y \in Y$ that can be reached from $Z \cup \{v\}$ via a trek whose left part only takes edges in G_{lat} and avoids H_2, H_1 .

Software

Algorithm. Recursively cycle through nodes $v \in \mathcal{O}$ to find LSC-tuples solving for $\bar{\Lambda}_{*, v}$.

- Find LSC-tuples via integer linear program that generalizes classical max-flow problem.
- Sound and complete.
- GitHub: [NilsSturma/LSC](#); soon on CRAN within SEMID package.

Example

1) The graph displayed on the left is LSC-identifiable.

$$\underline{v = v_1}: (Y, Z, H_1, H_2) = (\emptyset, \emptyset, \emptyset, \emptyset).$$

$$\underline{v = v_2, v_3, v_5, v_6}: (Y, Z, H_1, H_2) = (\{v_1\}, \emptyset, \emptyset, \emptyset) \text{ and } \Pi = \{v_1\}.$$

$$\underline{v = v_4}: (Y, Z, H_1, H_2) = (\{v_1, v_2, v_3\}, \{v_5\}, \emptyset, \{h_1\}) \text{ and}$$

$$\Pi = \{v_1, v_3, v_2 \leftarrow h_1 \rightarrow h_2 \rightarrow v_5\}.$$

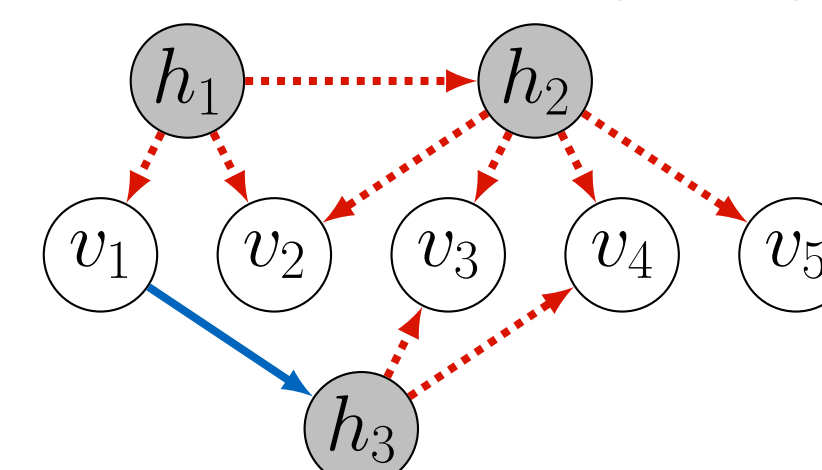
2) Now, consider the simpler model given by the **subgraph** G_{lat} with covariance matrix Ω . It is well known that $h_1 \rightarrow h_3$ is identifiable (up to sign) via

$$\sqrt{\frac{\omega_{v_2 v_6} \omega_{v_3 v_5}}{\omega_{v_2 v_3} \omega_{v_5 v_6} - \omega_{v_2 v_6} \omega_{v_3 v_5}}} = \sqrt{\lambda_{h_1 h_3}^2} = |\lambda_{h_1 h_3}|,$$

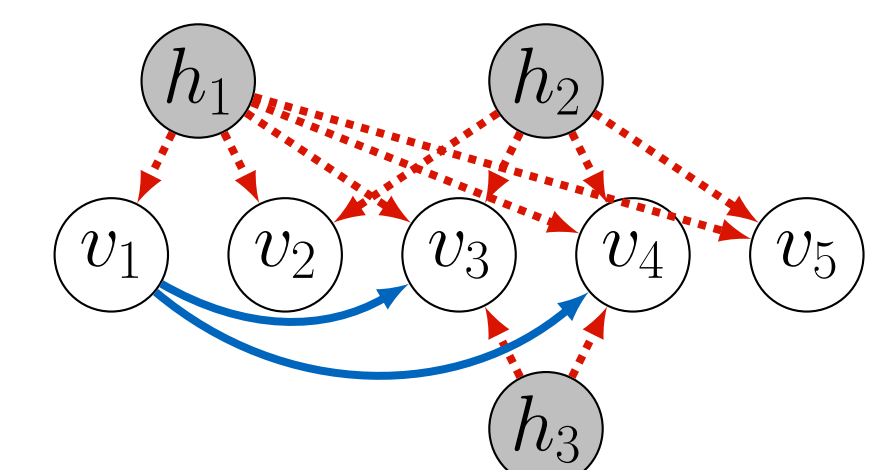
when fixing $\text{Var}[\varepsilon_h] = \phi_h = 1$ for all $h \in \mathcal{L}$.

Subtleties

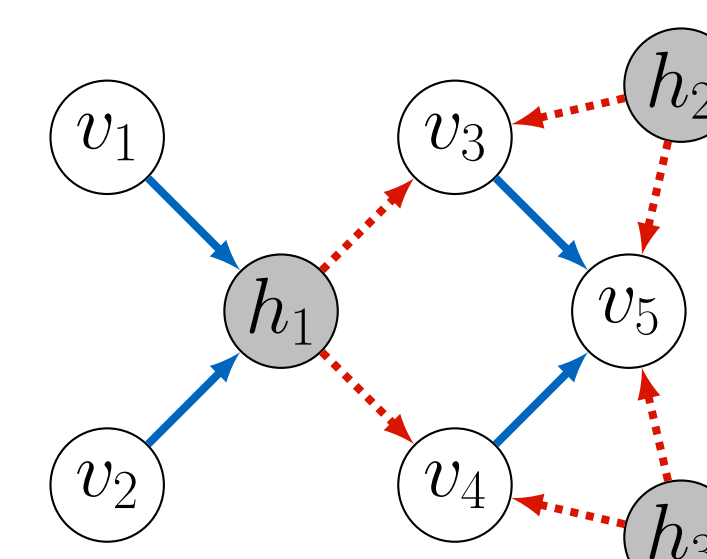
Canonicalization: Left side original graph, right side its canonicalization.



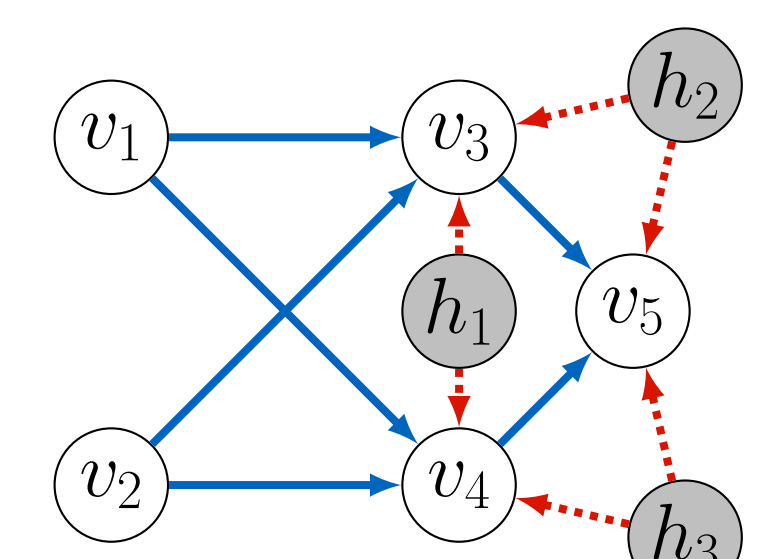
rationally identifiable



not rationally identifiable

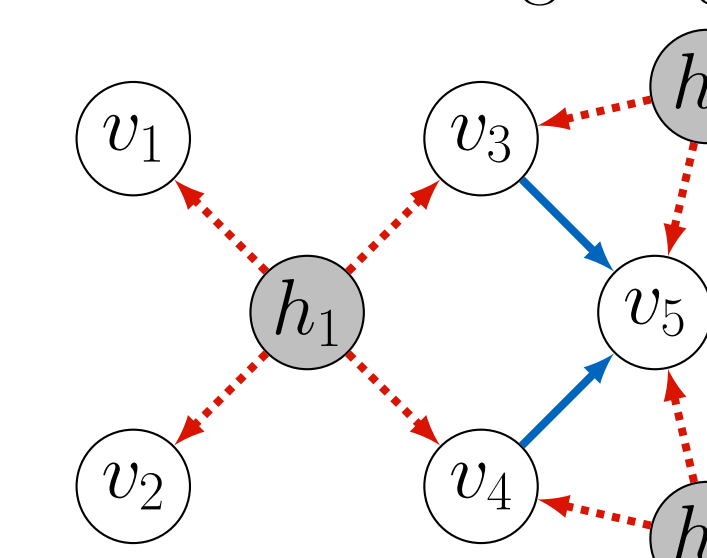


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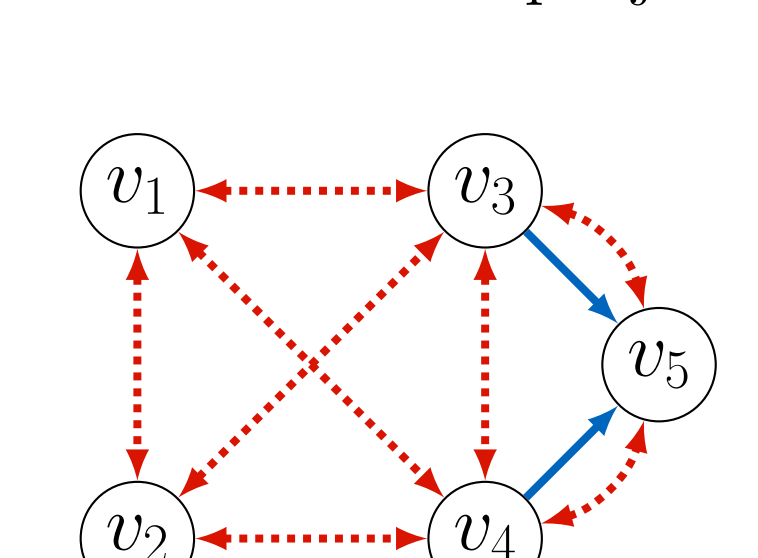


rationally identifiable

Latent projection: Left side original graph, right side its latent projection.



not rationally identifiable



rationally identifiable