

IDENTIFYING CAUSAL EFFECTS WITH COMPUTER ALGEBRA

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Parameter Identification In Graphical Models
The American Institute of Mathematics
October 4, 2010



STRUCTURAL EQUATION MODELS

- 1 The relationships among a set of observed variables are expressed by **linear equations**.
- 2 Each equation describes the dependence of one variable in terms of the others, and contains a **stochastic error term** accounting for the influence of unobserved factors.
- 3 Independence assumptions on pairs of error terms are also specified in the model.

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GAUSSIAN STRUCTURAL EQUATION MODELS

Let $G = (V, D, B)$ be a graph with vertex set $V = \{1, 2, \dots, m\}$, a set of **directed edges** D , and a set of **bidirected edges** B . Assume the subgraph of directed edges is acyclic and topologically ordered.

Let PD_n denote the set of $m \times m$ **symmetric positive definite matrices**.

Let $PD(B) := \{\Omega \in PD_m : \omega_{ij} = 0 \text{ if } i \neq j \text{ and } i \leftrightarrow j \notin B\}$.

Let $\epsilon \sim \mathcal{N}(0, \Omega)$ such that $\Omega \in PD(B)$.

For each $i \rightarrow j \in D$ let $\lambda_{ij} \in \mathbb{R}$ be a **parameter**. For each $j \in V$ define

$$X_j = \sum_{i:i \rightarrow j \in D} \lambda_{ij} X_i + \epsilon_j.$$

GAUSSIAN STRUCTURAL EQUATION MODELS

$$X_j = \sum_{i:i \rightarrow j \in D} \lambda_{ij} X_i + \epsilon_j, \text{ for } j \in V$$

Let Λ be the strictly upper triangular matrix with $\Lambda_{ij} = \lambda_{ij}$ if $i \rightarrow j \in D$ and $\Lambda_{ij} = 0$ otherwise.

Since D is an acyclic digraph, the random vector $X = (X_1, \dots, X_m)$ is well-defined centered multivariate normal distribution with **covariance matrix**

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

IDENTIFICATION PROBLEM

Decide whether the parameters in a structural model can be **determined uniquely** from the covariance matrix of the observed variables.

Equivalently, decide whether the following map is **injective**

$$\phi_G : (\Lambda, \Omega) \longrightarrow (I - \Lambda)^{-T} \Omega (I - \Lambda)^{-1}$$

The identification of a model is important because, in general, no reliable quantitative conclusion can be derived from a non-identified model.

EXAMPLE (PEARL 2000)

This model investigates the relations between **smoking** X and **lung cancer** Y , taking into consideration the **amount of tar** Z deposited in a person's lungs, and allowing for unobserved factors to affect both smoking X and cancer Y .

$$X = \varepsilon_1$$

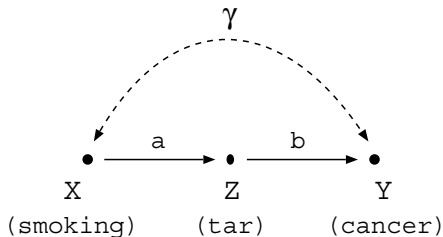
$$Z = aX + \varepsilon_2$$

$$Y = bZ + \varepsilon_3$$

$$\text{cov}(\varepsilon_1, \varepsilon_2) = 0$$

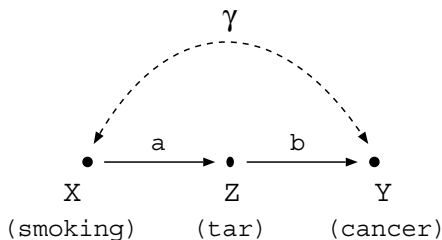
$$\text{cov}(\varepsilon_2, \varepsilon_3) = 0$$

$$\text{cov}(\varepsilon_1, \varepsilon_3) = \gamma$$



where $\varepsilon_i \sim \mathcal{N}(0, \omega_i)$.

EXAMPLE (PEARL 2000)

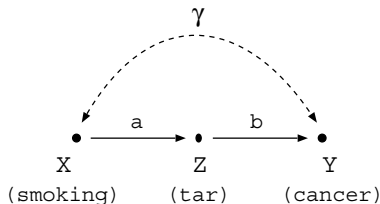


$$\Sigma = (I - \Lambda)^{-T} \Omega (I - \Lambda)^{-1}$$

$$\begin{bmatrix} \sigma_{11} & \sigma_{12} & \sigma_{13} \\ \sigma_{12} & \sigma_{22} & \sigma_{23} \\ \sigma_{13} & \sigma_{23} & \sigma_{33} \end{bmatrix} = \begin{bmatrix} \omega_1 & a\omega_1 & ab\omega_1 + \gamma \\ a\omega_1 & a^2\omega_1 + \omega_2 & a^2b\omega_1 + b\omega_2 + a\gamma \\ ab\omega_1 + \gamma & a^2b\omega_1 + b\omega_2 + a\gamma & a^2b^2\omega_1 + b^2\omega_2 + \omega_3 + 2ab\gamma \end{bmatrix}$$

EXAMPLE (PEARL 2000)

$$\begin{bmatrix} \sigma_{11} & \sigma_{12} & \sigma_{13} \\ \sigma_{12} & \sigma_{22} & \sigma_{23} \\ \sigma_{13} & \sigma_{23} & \sigma_{33} \end{bmatrix}$$

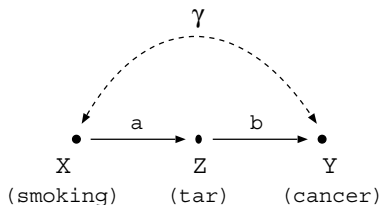


$$a = \frac{\sigma_{12}}{\sigma_{11}} \quad b = \frac{\sigma_{12}\sigma_{13} - \sigma_{11}\sigma_{23}}{\sigma_{12}^2 - \sigma_{11}\sigma_{22}}$$

$$\omega_1 = \sigma_{11} \quad \omega_2 = \frac{\sigma_{11}\sigma_{22} - \sigma_{12}^2}{\sigma_{11}} \quad \gamma = \frac{\sigma_{11}\sigma_{12}\sigma_{23} - \sigma_{11}\sigma_{13}\sigma_{22}}{\sigma_{12}^2 - \sigma_{11}\sigma_{22}}$$

$$\omega_3 = \frac{1}{(\sigma_{12}^2 - \sigma_{11}\sigma_{22})^2} (\sigma_{12}^2\sigma_{13}^2\sigma_{22} - 2\sigma_{12}^3\sigma_{13}\sigma_{23} + 2\sigma_{11}\sigma_{12}^2\sigma_{23}^2 - \sigma_{11}^2\sigma_{22}\sigma_{23}^2 + \sigma_{12}^4\sigma_{33} - 2\sigma_{11}\sigma_{12}^2\sigma_{22}\sigma_{33} + \sigma_{11}^2\sigma_{22}^2\sigma_{33})$$

EXAMPLE (PEARL 2000)



$$\begin{bmatrix} \sigma_{11} & \sigma_{12} & \sigma_{13} \\ \sigma_{12} & \sigma_{22} & \sigma_{23} \\ \sigma_{13} & \sigma_{23} & \sigma_{33} \end{bmatrix} = \begin{bmatrix} \omega_1 & a\omega_1 & ab\omega_1 + \gamma \\ a\omega_1 & a^2\omega_1 + \omega_2 & a^2b\omega_1 + b\omega_2 + a\gamma \\ ab\omega_1 + \gamma & a^2b\omega_1 + b\omega_2 + a\gamma & a^2b^2\omega_1 + b^2\omega_2 + \omega_3 + 2ab\gamma \end{bmatrix}$$

ϕ_G IS GLOBALLY IDENTIFIABLE

The map $\phi_G : (\Lambda, \Omega) \rightarrow (I - \Lambda)^{-T} \Omega (I - \Lambda)^{-1}$ is injective **on the entire possible domain of definition.**

A directed graph D with at least two nodes is an **arborescence** converging to node i if i is the unique **sink** of D .

Given a mixed graph $G = (V, D, B)$ and a subset $A \subset V$, the **mixed subgraph induced by A** , denoted by $G_A = (A, D_A, B_A)$ is the graph containing all directed and bidirected edges whose endpoints are in A .

THEOREM (DRTON, FOYCEL, SULLIVANT)

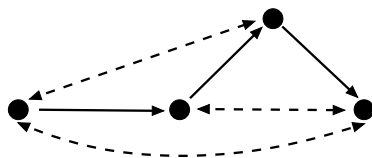
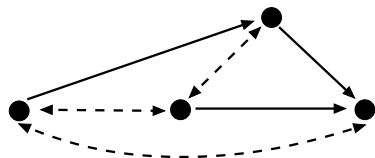
The parametrization ϕ_G fails to be injective if and only if there is an induced subgraph G_A whose directed part (A, D_A) contains a converging arborescence and whose bidirected part (A, B_A) is connected.

GLOBAL IDENTIFIABILITY

The mixed graph $G = (V, D, B)$ is **simple**, or **bow-free** if $D \cap B = \emptyset$.

COROLLARY

Suppose the map ϕ_G given by an acyclic mixed graph G is **injective**. Then G is a **simple**.



The two unlabeled simple graphs on four nodes with non-injective parametrization.

GENERICALLY IDENTIFIABLE

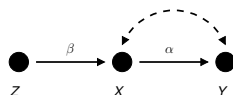
THEOREM (BRITO, PERL)

Any simple graph G is generically identifiable.

$$\alpha = \frac{\sigma_{13}}{\sigma_{12}}$$

$$\beta = \frac{\sigma_{12}}{\sigma_{11}}$$

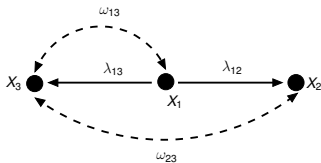
$$\omega_{XY} = \frac{\sigma_{12}\sigma_{23} - \sigma_{13}\sigma_{22}}{\sigma_{12}}$$



The parameter α is identified as long as $\sigma_{12} \neq 0$. The **instrumental variable** graph is not globally identified but **generically identifiable**.

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

So if the parameters in Λ are generically identifiable then the parameters in Ω are generically identifiable.



The parameter ω_{23} is identified, but λ_{13} is not identified.

$$\omega_{23} = \frac{\sigma_{11}\sigma_{23} - \sigma_{12}\sigma_{13}}{\sigma_{11}}$$

APPROACHES TO THE IDENTIFICATION PROBLEM

Algebraic manipulation of the equations defining the model.

- 1 The method of path coefficients (Wright, 1934)
- 2 The rank and order criteria (Fisher, 1966)
- 3 Block recursive models (Fisher, 1966; Rigdon 1995)

Graphical Methods.

- 1 Single door criterion (Pearl, 2000)
- 2 Instrumental variables (Bowden and Turkington, 1984)
- 3 Back door criterion for total effects (Pearl, 2000)
- 4 G-criterion (Brito, 2006)
- 5 Graphical methods introduced by Tian (2004; 2005; 2007; 2009)
- 6 Recanting witness criterion for path-specific effects (Avin, Shpitser and Pearl, 2005)

It remains unclear if these criteria (or combinations of the criteria) are **necessary and sufficient** to decide whether or not parameters are generically identifiable in a general mixed graph.

Introduce a **general** algebraic framework for performing identifiability computations for graphical models.

- Capable of testing direct effects, total effects, path-specific effects, error variances and covariances (Ω parameters).
- Provides certificates if a given parameter is not identifiable.
- Capable of detecting **algebraic** d -identifiable parameters (d -to-one parameters).

BASIC COMPUTATIONAL ALGEBRAIC GEOMETRY

Let k be a field. An **affine variety** is the common zero locus of polynomials $f_1, \dots, f_r \in k[x_1, \dots, x_n]$.

$$V(f_1, \dots, f_r) = \{f_1 = 0, f_2 = 0, \dots, f_r = 0\}$$

$$V(1) = \emptyset$$

$$V(0) = k^n$$

For **linear polynomials** f_i , $V(f_1, \dots, f_r)$ is the solution space of an inhomogeneous system of linear equations. This variety is described parametrically applying **Gauss Algorithm**.

Given f_1, \dots, f_r in $k[x_1, \dots, x_n]$,

$$\langle f_1, \dots, f_r \rangle = \left\{ \sum_{i=1}^n h_i f_i \mid h_1, \dots, h_n \in k[x_1, \dots, x_n] \right\}$$

is the **ideal** generated by f_1, \dots, f_r .

THEOREM (HILBERT BASIS THEOREM)

If $I \subset k[x_1, \dots, x_n]$ is an ideal, there exists f_1, \dots, f_r such that

$$\langle f_1, \dots, f_r \rangle = I.$$

RATIONAL PARAMETRIZATION

Let $p_1(t_1, \dots, t_d), \dots, p_n(t_1, \dots, t_d) \in k[t_1, \dots, t_d]$. The set

$$S = \{(p_1(a_1, \dots, a_d), \dots, p_n(a_1, \dots, a_d)) \in k^n \mid (a_1, \dots, a_d) \in k^d\}$$

is called a **rational parametrization (r.p.)**.

$$I(S) = \{g \in k[x_1, \dots, x_n] \mid g(a_1, \dots, a_n) = 0 \text{ for all } (a_1, \dots, a_n) \in S\}$$

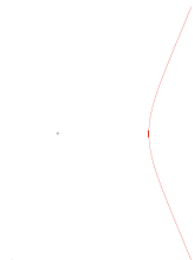
$I(S)$ is the **ideal of polynomial functions vanishing on S** .

THEOREM

If S is a r.p. and $I(S) = \langle f_1, \dots, f_r \rangle$ then S and $V(f_1, \dots, f_r)$ differ by a set of dimension less than the dimension of S .

CUBIC PLANE CURVE

Let $S = \{(t^2 + 1, t^3 + t) \mid t \in \mathbb{R}\}$. Then $I(S) = \langle y^2 - x^3 - x^2 \rangle$



PROJECTION AND ELIMINATION

For any ideal $I \subset k[x_1, \dots, x_n]$ we consider the **elimination ideal**

$$I_m = I \cap k[x_{m+1}, \dots, x_n]$$

and the **projection** $\pi_m : k^n \rightarrow k^{n-m}$ given by

$$\pi_m(a_1, \dots, a_n) = (a_{m+1}, \dots, a_n)$$

THEOREM

If k is **algebraically closed**, then $\overline{\pi_m(V(I))} = V(I_m)$.

THEOREM

If $G = \{g_1, \dots, g_r\}$ is a **Gröbner basis** of $I \subset k[x_1, \dots, x_n]$ with respect to the lexicographic ordering, then

$$G_m = G \cap k[x_{m+1}, \dots, x_n]$$

is a Gröbner basis of $I_m \subset k[x_{m+1}, \dots, x_n]$ w.r.t. *lex*.

THEOREM

Suppose k is an **infinite field** and we are given a rational map $\phi : k^m \setminus Z \rightarrow k^n$ given by

$$(t_1, \dots, t_m) \mapsto \left(\frac{f_1(t)}{g_1(t)}, \dots, \frac{f_n(t)}{g_n(t)} \right)$$

with f_i and $g_i \in k[t_1, \dots, t_m]$ and $Z = V(g_1 \cdot g_2 \cdots g_n)$. Let

$$I = \langle g_1 x_1 - f_1, \dots, g_n x_n - f_n, 1 - g_s \rangle \subset k[s, t_1, \dots, t_m, x_1, \dots, x_n]$$

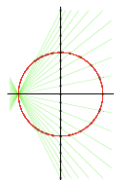
Then

$$\overline{\text{Im}(\phi)} = V(I_{m+1})$$

IMPLICITIZATION OF THE UNIT CIRCLE

Consider the parametrization of the unit circle given by the 2nd intersection point of $y = t(x + 1)$ with the circle:

$$t \mapsto \left(\frac{1 - t^2}{t^2 + 1}, \frac{2t}{t^2 + 1} \right)$$



Let $I = \langle (t^2 + 1)x - (1 - t^2), (t^2 + 1)y - 2t, 1 - (t^2 + 1)^2s \rangle$. The set

$$\left\{ x^2 + y^2 - 1, ty + x - 1, tx + t - y, s - \frac{1}{2}x - \frac{1}{2} \right\}$$

is a Gröbner basis with respect to $\text{lex } s > t > x > y$. Hence

$$I_2 = \langle x^2 + y^2 - 1 \rangle$$

.

IDENTIFIABLE PARAMETERS

Let $\Theta \subseteq \mathbb{R}^d$ be a full dimensional **parameter set**.

Let $\phi_1, \dots, \phi_n \in \mathbb{R}[t_1, \dots, t_d]$, and $\Phi : \Theta \rightarrow \mathbb{R}^n$ be the function defined by $\Phi(\theta) = (\phi_1(\theta), \dots, \phi_n(\theta))^T$.

A **parameter** is a polynomial function $u : \Theta \rightarrow \mathbb{R}$ which is not constant on Θ .

The parameter u is **identifiable** if there exists a map $\Psi : \mathbb{R}^n \rightarrow \mathbb{R}$ such that $u(\theta) = \Psi \circ \Phi(\theta)$ for all $\theta \in \Theta$.

The parameter u is **generically identifiable** if there exists a map $\Psi : \mathbb{R}^n \rightarrow \mathbb{R}$ and a dense open subset \mathcal{O} of Θ such that $u(\theta) = \Psi \circ \Phi(\theta)$ for all $\theta \in \mathcal{O}$.

Given $\Phi : \Theta \rightarrow \mathbb{R}^n$ defined by $\Phi(\theta) = (\phi_1(\theta), \dots, \phi_n(\theta))^T$, we want to check if the **parameter** u is (generically) **identifiable**.

The **vanishing ideal** of $S \subseteq \mathbb{R}^n$ is the set

$$\mathcal{I}(S) := \{g \in \mathbb{R}[p_1, \dots, p_n] : g(\mathbf{a}) = 0 \text{ for all } \mathbf{a} \in S\}.$$

Let $\tilde{\Phi} = (u, \phi_1, \dots, \phi_n)^T : \Theta \rightarrow \mathbb{R}^{d+1}$.

Let $\mathbb{R}[q, \mathbf{p}]$ be the polynomial ring with **one extra indeterminate** corresponding to the parameter function u .

Let $\mathcal{I}(\tilde{\Phi}(\Theta))$ be the vanishing ideal of the image.

THEOREM

Suppose that $g(q, \mathbf{p}) \in \mathcal{I}(\tilde{\Phi}(\Theta))$ is a polynomial such that q appears in this polynomial, $g(q, \mathbf{p}) = \sum_{i=0}^d g_i(\mathbf{p})q^i$ and $g_d(\mathbf{p})$ does not belong to $\mathcal{I}(\Phi(\Theta))$.

- 1 If g is **linear** in q , $g = g_1(\mathbf{p})q - g_0(\mathbf{p})$ then u is **generically identifiable** by the formula $u = \frac{g_0(\mathbf{p})}{g_1(\mathbf{p})}$. If, in addition, $g_1(\mathbf{p}) \neq 0$ for $\mathbf{p} \in \Phi(\Theta)$ then u is **identifiable**.
- 2 If g has **higher degree** d in q , then u is **algebraically d -identifiable** (may or might not be identifiable).
- 3 If no such polynomial g exists then the parameter u is **not generically identifiable**.

THEOREM

Of the 64 mixed graphs on three vertices,

- *there are exactly 31 graphs that are **generically identifiable** and 33 graphs that are **not generically identifiable**.*
- *The single-door criterion and instrumental variables form a **complete method** to generically identify direct causal effects for SEM models on three variables.*

THEOREM

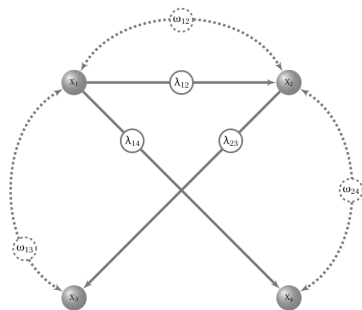
Of the 4096 mixed graphs on four variables

- *exactly 1246 are **generically identifiable**, 6 are **algebraically 2-identified**, and 2844 are **not generically identifiable**.*
- *Of the 1246 generically identifiable models, exactly 1093 are generically identified by the single-door and instrumental variables criteria and the remaining 153 generically identified models contain direct causal effect parameters **only identified by the algebraic method**.*
- *There are exactly 729 bow-free models, each generically identified by the single-door criterion.*

<http://www.shsu.edu/research/graphicalmodels/>

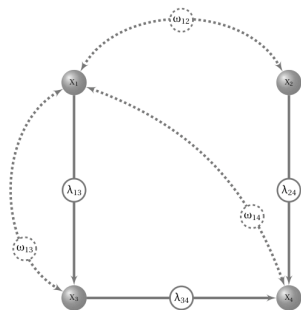
CHALLENGE 1

- Show that λ_{12} is generically identifiable.
- Do you know of a graphical criterion that can identify this parameter?
- Show that G is generically identifiable.



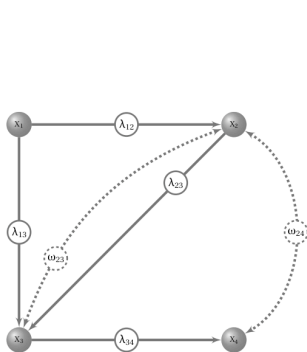
CHALLENGE 2

- Show that λ_{24} and λ_{34} are generically identifiable.
- Do you know of a graphical criterion that can identify these parameters?
- Show that G is generically identifiable.



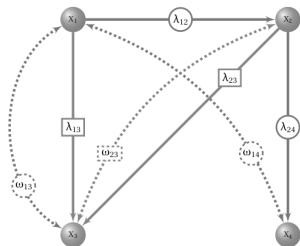
CHALLENGE 3

- Show that λ_{13} and λ_{23} are generically identifiable.
- Do you know of a graphical criterion that can identify these parameters?
- Show that G is generically identifiable.



CHALLENGE 4

- Show that λ_{13} and λ_{23} are **not** generically identifiable.
- Show that the **total effect** of X_1 on X_3 : $\lambda_{12}\lambda_{23} + \lambda_{13}$ is generically identifiable.
- This total effect is **not** identified by the **Back-Door criterion**. Is there a graphical criterion that identifies this parameter?



CHALLENGE 5

TABLE: Algebraically 2-identified SEMs on four variables.

Directed edges	Bidirected edges
$1 \rightarrow 2, 2 \rightarrow 3, 3 \rightarrow 4$	$1 \leftrightarrow 2, 1 \leftrightarrow 3, 1 \leftrightarrow 4$
$1 \rightarrow 2, 2 \rightarrow 3, 2 \rightarrow 4$	$1 \leftrightarrow 2, 1 \leftrightarrow 3, 1 \leftrightarrow 4$
$1 \rightarrow 2, 1 \rightarrow 4, 2 \rightarrow 3$	$1 \leftrightarrow 2, 1 \leftrightarrow 3, 1 \leftrightarrow 4$
$1 \rightarrow 2, 1 \rightarrow 3, 3 \rightarrow 4$	$1 \leftrightarrow 2, 1 \leftrightarrow 3, 1 \leftrightarrow 4$
$1 \rightarrow 2, 1 \rightarrow 3, 2 \rightarrow 4$	$1 \leftrightarrow 2, 1 \leftrightarrow 3, 1 \leftrightarrow 4$
$1 \rightarrow 2, 1 \rightarrow 3, 1 \rightarrow 4$	$1 \leftrightarrow 2, 1 \leftrightarrow 3, 1 \leftrightarrow 4$

Is there a combinatorial description of these mixed graphs?