

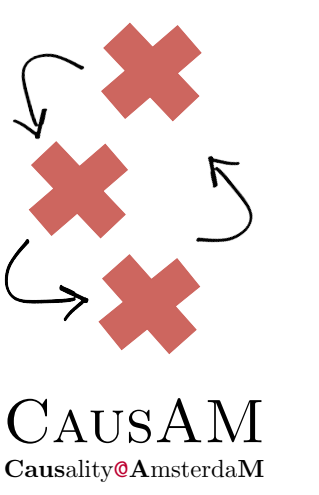
Curing the Curse of Non-Recursiveness in Structural Causal Models

Stephan Bongers¹, Jonas Peters², Bernhard Schölkopf³, Joris M. Mooij¹

¹: University of Amsterdam;
The Netherlands

²: University of Copenhagen;
Denmark

³: Max Planck Institute for Intelligent Systems;
Germany



Abstract

In this work (Bongers et al., 2016) we give a general treatment of **structural causal models (SCMs)** in the **cyclic** setting. We show that if one is only interested in a particular subset of endogenous variables, one can under certain conditions arrive at a more parsimonious representation of the SCM by means of **marginalizing** out endogenous and **reducing** exogenous variables.

Structural Causal Models

$$\text{SCM} = \begin{cases} \text{Structural equations:} & \mathbf{x} = \mathbf{f}(\mathbf{x}, \mathbf{e}), \\ \text{Exogenous distribution:} & \mathbb{P}_{\mathcal{E}}. \end{cases}$$

Example:

$$\begin{aligned} f_u(\mathbf{x}, \mathbf{e}) &= e_b + e_c \\ f_v(\mathbf{x}, \mathbf{e}) &= x_u + x_j + e_d \\ p(e_b, e_c, e_d) &= \mathcal{N}(\mathbf{0}, \mathbf{I}) \end{aligned}$$

An SCM is **uniquely solvable** if the structural equations have a unique solution. Every acyclic SCM is uniquely solvable. **Extending the above model to the SCM \mathcal{M} :**

$$\begin{aligned} f_h(\mathbf{x}, \mathbf{e}) &= x_u + x_j + e_a \\ f_i(\mathbf{x}, \mathbf{e}) &= x_j + e_b \\ f_j(\mathbf{x}, \mathbf{e}) &= \alpha x_i \end{aligned}$$

leads to a non-uniquely solvable SCM for $\alpha = 1$.

Unique solvability implies that the SCM has a **solution** in terms of random variables (\mathbf{X}, \mathbf{E}) :

- (1) $\mathbb{P}^{\mathbf{E}} = \mathbb{P}_{\mathcal{E}}$
- (2) $\mathbf{X} = \mathbf{f}(\mathbf{X}, \mathbf{E})$ a.s..

that has a unique distribution.

Perfect interventions modelled à la Pearl. The **perfect intervention** $\text{do}(j, 5)$ on \mathcal{M} changes the causal mechanism of f_j to

$$f_j(\mathbf{x}, \mathbf{e}) = 5$$

which yields a uniquely solvable SCM $\mathcal{M}_{\text{do}(j,5)}$.

Cyclic SCMs are challenging, since unique solvability is not guaranteed (not even preserved under perfect interventions).

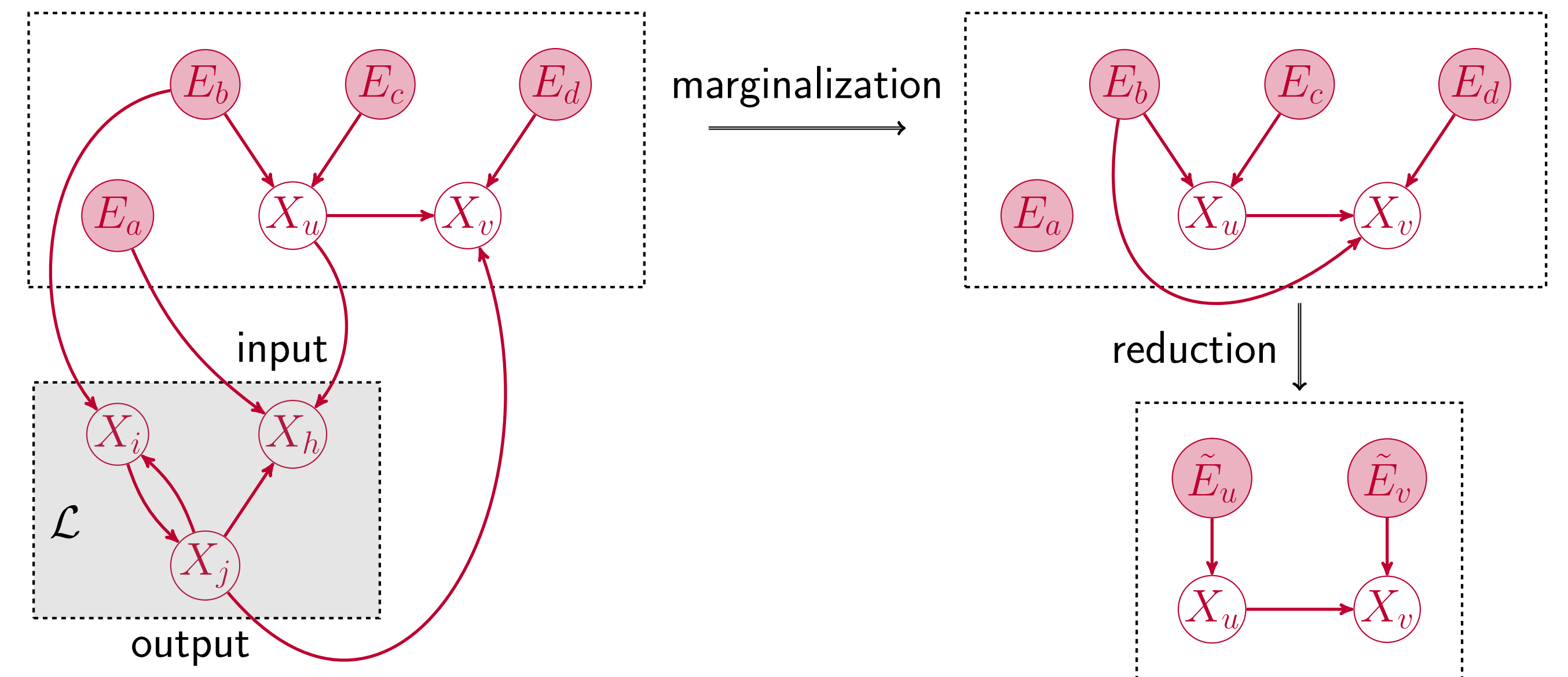
Marginalization

Define the **marginalization** $\mathcal{M}_{\text{marg}(\mathcal{L})}$ over the subset of endogenous variables \mathcal{L} by

$$\mathbf{f}_{\text{marg}(\mathcal{L})}(\mathbf{x}, \mathbf{e}) := \mathbf{f}(\mathbf{g}_{\mathcal{L}}(\mathbf{x}, \mathbf{e}), \mathbf{x}, \mathbf{e})$$

where $\mathbf{g}_{\mathcal{L}}$ is the induced mapping obtained by solving the structural equations of $\mathbf{f}_{\mathcal{L}}$ for the variables \mathcal{L} . For $\alpha = \frac{1}{2}$ this gives

$$\begin{aligned} x_h &= x_u + e_a + e_b & g_h(\mathbf{x}, \mathbf{e}) &= x_u + e_a + e_b \\ x_i &= 2e_b & \implies g_i(\mathbf{x}, \mathbf{e}) &= 2e_b \\ x_j &= e_b & g_j(\mathbf{x}, \mathbf{e}) &= e_b \end{aligned}$$



Theorem 1 Let \mathcal{M} be an SCM that is uniquely solvable w.r.t. \mathcal{L} . Then \mathcal{M} and its marginalization $\mathcal{M}_{\text{marg}(\mathcal{L})}$ are interventionally equivalent w.r.t. the complement of \mathcal{L} .

Exogenous Reparametrization

Define the **exogenous reparametrization** $\mathcal{M}_{\text{rep}(\phi)}$ with respect to a reparameterization ϕ by $\mathbf{f}_{\text{rep}(\phi)}$ such that

$$\mathbf{f}_{\text{red}(\phi)}(\mathbf{x}, \phi(\mathbf{e})) = \mathbf{f}(\mathbf{x}, \mathbf{e}).$$

In the example, this gives:

$$\begin{aligned} \phi_u(\mathbf{e}) &= e_b + e_c & \implies f_{\text{rep}(\phi),u}(\mathbf{x}, \tilde{\mathbf{e}}) &= \tilde{e}_u \\ \phi_v(\mathbf{e}) &= e_b + e_d & \implies f_{\text{rep}(\phi),v}(\mathbf{x}, \tilde{\mathbf{e}}) &= x_u + \tilde{e}_v \end{aligned}$$

Theorem 2 An exogenously reparametrized SCM is interventionally equivalent to the original SCM.

Corollary 1 Every real-valued SCM \mathcal{M} can be exogenously reparametrized to a real-valued SCM $\mathcal{M}_{\text{rep}(\phi)}$ with only a single one-dimensional real-valued exogenous variable.

Reduction

A **reduction** \mathcal{M}_{red} is an interventionally equivalent SCM with a smaller space of exogenous variables.

By Corollary 1, **reductions** always exist, however \mathbf{f}_{red} is typically very wild (hard to estimate from data!).

We introduce a class of **nice reductions** that generalize smooth reductions of linear models and Markovian models.

Theorem 3 Nice reductions do not exist in general.

For example, no nice reduction exists if we change f_v to:

$$\begin{aligned} f_v(\mathbf{x}, \mathbf{e}) &= x_u x_j + e_d & \implies f_{\text{marg}(\mathcal{L}),u}(\mathbf{x}, \mathbf{e}) &= e_b + e_c \\ & & f_{\text{marg}(\mathcal{L}),v}(\mathbf{x}, \mathbf{e}) &= x_u e_b + e_d \end{aligned}$$

Discussion

What parameterisations of SCMs should we use, given that nice reductions do not exist?

References

Bongers, S., Peters, J., Schölkopf, B., and Mooij, J. M. (2016). Structural causal models: Cycles, marginalizations, exogenous reparametrizations and reductions. *arXiv.org preprint*, arXiv:1611.06221 [stat.ME].

SB and JMM were supported by NWO (VIDI grant 639.072.410 and VENI grant 639.031.036). JMW was also supported by ERC (grant agreement n° 639466).