

Definition 1 (Unique Solvability). *Let $M = \langle V, U, F, P(U) \rangle$ be an SCM and $\mathbf{Z} \subseteq \mathbf{V}$ a subset of endogenous variables. We say that M is uniquely solvable if for almost every $\mathbf{u} \in \text{dom}(\mathbf{U})$ the equations*

$$\mathbf{V} = F(\mathbf{V}, \mathbf{u})$$

have a unique solution.

Throughout this paper we only need consider uniquely solvable SCMs, which simplifies our definition of potential response.

Definition 2 (Potential Response). *Let $M = \langle V, U, F, P(U) \rangle$ be a uniquely solvable SCM. The potential response function is the mapping $\bar{\mathbf{V}}(\mathbf{U}) : \text{dom}(\mathbf{U}) \rightarrow \text{dom}(\mathbf{V})$ which associates each $\mathbf{u} \in \text{dom}(\mathbf{U})$ with the unique solution \mathbf{v}^* of $\mathbf{V} = F(\mathbf{V}, \mathbf{u})$.*

Definition 3 (Simple SCM). *Let $M = \langle V, U, F, P(U) \rangle$ be an SCM. We call M simple if M is uniquely solvable w.r.t. every subset $\mathbf{Z} \subseteq \mathbf{V}$.*

The following definition is adopted from [?].

Definition 4 (Lipschitz Matrix). *Let $M = \langle V, U, F, P(U) \rangle$ be an SCM, with $\text{dom}(\mathbf{U}) = \mathbb{R}^m$, $\text{dom}(\mathbf{V}) = \mathbb{R}^n$, and $F : \text{dom}(\mathbf{U}) \times \text{dom}(\mathbf{V}) \rightarrow \text{dom}(\mathbf{V})$ differentiable and Lipschitz.*

Let $\mathbf{Z} \subseteq \mathbf{V}$ be a subset of endogenous variables. For each pair of vertex indices $i, j \in \mathbf{Z}$, define

$$a_{ij} = \sup_{\mathbf{u}, \mathbf{v}} \left| \frac{\partial f_i}{\partial v_j}(\mathbf{u}, \mathbf{v}) \right|$$

We call the matrix $A_{\mathbf{Z}} = [a_{ij}]_{i,j \in \mathbf{Z}}$ the Lipschitz matrix of $F_{\mathbf{Z}}$.

When $\mathbf{Z} = \mathbf{V}$, we simply call $A = [a_{ij}]$ the Lipschitz matrix of F .

The matrix A can be viewed as a collection of Lipschitz constants along the direction of each partial derivative.

Definition 5 (stable-Lipschitz SCM). *Let $M = \langle V, U, F, P(U) \rangle$ be an SCM with causal diagram G . We say M is stable-Lipschitz if, for every strongly connected component \mathbf{Z} of G , the following conditions hold:*

- $F_{\mathbf{Z}}$ is differentiable and Lipschitz.
- $\rho(A_{\mathbf{Z}}) < 1$.

No constraints are placed on components of F which are not part of strongly connected components of G . This immediately implies the following:

Theorem 1 (acyclic \subset stable-Lipschitz). *Let M be an acyclic SCM. Then M is stable-Lipschitz.*

Theorem 2 (Closed under Interventions). *Let M be a stable-Lipschitz SCM, $\mathbf{X} \subseteq \mathbf{V}$, and $\mathbf{x} \in \text{dom}(\mathbf{X})$. Then $M_{\text{do}(\mathbf{X}=\mathbf{x})}$ is stable-Lipschitz.*

Lemma 3 (Unique Solvability). *Let M be a stable-Lipschitz SCM. Then M is uniquely solvable.*

Theorem 4 (stable-Lipschitz \subset simple). *Let M be a stable-Lipschitz SCM. Then M is simple.*

New Results

Lemma 5 (Injectivity). *Let M have structural functions of the form $F(\mathbf{V}, \mathbf{U}) = H(\mathbf{V}) + \mathbf{U}$ (additive noise). Furthermore, assume that for every ancestral $\mathbf{W} \subseteq \mathbf{V}$,*

- $F_{\mathbf{W}}$ is Lipschitz
- $\det(I_{|\mathbf{W}|} - A_{\mathbf{W}}) \neq 0$ (uniquely solvable)

Then $h_{\mathbf{V}} := \mathbf{V} - H(\text{Pa}(\mathbf{V}))$ is injective for ancestral $\mathbf{W} \subseteq \mathbf{V}$.

Lemma 6 (Intrinsic Stability). *Let A be a Lipschitz matrix for differentiable $H(\mathbf{V})$. Then $\rho(A) < 1$ implies $\rho(\frac{dK}{d\mathbf{V}}(\mathbf{v})) < 1$ for all $\mathbf{v} \in \mathbf{V}$.*

Theorem 7 (Observational dGMP). *Let M be stable Lipschitz with*
1. *structural equations of the form $F(\mathbf{V}, \mathbf{U}) = H(\mathbf{V}) + \mathbf{U}$ (additive noise),*
2. *each $U_i \cap U_j = \emptyset$ for $i \neq j$ (independent noise), and*
3. *$P_M(\mathbf{V})$ has density according to the Legesgue measure on $\mathbb{R}^{|\mathbf{V}|}$ (positivity).*

Then M satisfies the directed global Markov property.

Remark 1. *I am confident that conditions 1 and 2 in the hypothesis can be substantially weakened with further research.*

This is a very new result, so while I believe the proof to be accurate and comprehensive, I'm still vetting it for errors: I'd place 5:1 odds against finding an irrecoverable error in the proof.

Corollary 1 (Adjustment Formula). *Let M be as (before) and $Q = P(y|\text{do}(x))$ a causal query. If the BDC is satisfied, then Q can be found via backdoor adjustment.*

Remark 2. *One of the motivations for weakening the condition of independent noise is to be able to similarly prove that the front-door criteria is also valid.*

Theorem 8 (Interventional dGMP). *Let M be as in Theorem 7. For any $\mathbf{X} \subseteq \mathbf{Z}$, $M_{\text{do}(\mathbf{X}=\mathbf{x})}$ satisfies the directed global Markov property.*

Theorem 9 (Closure under Twin Operation). *Let M be stable-Lipschitz. Then M^{twin} is also stable-Lipschitz.*

Conjecture 1 (Counterfactual dGMP). *Let M be stable-Lipschitz. Then the counterfactual distributions of M satisfy the directed global Markov property relative to the corresponding twin network.*

Remark 3. *I believe the counterfactual dGMP holds if the condition of independent noise can be weakened, as an immediate consequence of the other results so far. However, I would place 2:1 odds that I'm missing some additional aspect of the proof.*