Appendix

A PROOFS FOR SECTION 4.3: ELEMENTARY EXACT TRANSFORMATIONS

Lemma 4. The identity mapping and permuting the labels of variables are both exact transformations. That is, if \mathcal{M}_X is an SEM and $\pi : \mathbb{I}_X \to \mathbb{I}_X$ is a bijection then the transformation

$$\tau : \mathcal{X} \to \mathcal{Y}$$
$$(x_i : i \in \mathbb{I}_X) \mapsto (x_{\pi(i)} : i \in \mathbb{I}_X)$$

naturally gives rise to an SEM \mathcal{M}_Y that is an exact τ -transformation of \mathcal{M}_X , corresponding to relabelling the variables.

Proof of Lemma 4. Consider the SEM \mathcal{M}_Y obtained from \mathcal{M}_X by replacing, for all $i \in \mathbb{I}_X$, any occurrence of X_i in the structural equations \mathcal{S}_X and interventions \mathcal{I}_X by $Y_{\pi(i)}$ and leaving the distribution over the exogenous variables unchanged.

Proof of Lemma 5 (Transitivity of exact transformations). Let $\omega_{ZY} : \mathcal{I}_Y \to \mathcal{I}_Z$ and $\omega_{YX} : \mathcal{I}_X \to \mathcal{I}_Y$ be the mappings between interventions corresponding to the exact transformations τ_{ZY} and τ_{YX} respectively and define $\omega_{ZX} = \omega_{ZY} \circ \omega_{YX} : \mathcal{I}_X \to \mathcal{I}_Z$. Then ω_{ZX} is surjective and order-preserving since both ω_{ZY} and ω_{YX} are surjective and order-preserving. Since τ_{ZY} and τ_{YX} are exact it follows that for all $i \in \mathcal{I}_X$

$$\mathbb{P}^{i}_{\tau_{ZX}(X)} = \mathbb{P}^{\omega_{ZY}(\omega_{YX}(i))}_{\tau_{ZY}(\tau_{YX}(X))} = \mathbb{P}^{\operatorname{do}(\omega_{ZX}(i))}_{Z}$$

i.e. \mathcal{M}_Z is an τ_{ZX} -exact transformation of \mathcal{M}_X .

B PROOFS FOR SECTION 5.1: MARGINALISATION OF VARIABLES

Proof of Theorem 9 (Marginalisation of childless variables). By Lemma 5 it suffices to proof this for marginalisation of one childless variable. Without loss of generality, let X_1 be the childless variable to be marginalised out.

Let $\mathcal{M}_Y = (\mathcal{S}_Y, \mathcal{I}_Y, \mathbb{P}_F)$ be the SEM where

- the structural equations S_Y are obtained from S_X by removing the structural equation corresponding to the childless variable X_1 ;
- \mathcal{I}_Y is the image of the map ω : $\mathcal{I}_X \to \mathcal{I}_Y$ that drops any reference to the variable X_1 (e.g. $do(X_1 = x_1, X_2 = x_2) \in \mathcal{I}_X$ would be mapped to $do(X_2 = x_2) \in \mathcal{I}_Y$);
- $F = (E_i : i \in \mathbb{I}_X \setminus \{1\})$ are the remaining noise variables distributed according to their marginal distribution under \mathbb{P}_E .

By construction, ω is surjective and order-preserving. Let $i \in \mathcal{I}_X$ be any intervention. The variable X_1 being childless ensures that the law on the remaining variables $X_k, k \in \mathbb{I}_X \setminus \{1\}$ that we obtain by *marginalisation* of the childless variable, i. e. $\mathbb{P}^i_{\tau(X)}$, is equivalent to the law one obtains by simply *dropping* the childless variable, which is exactly what the law under \mathcal{M}_Y amounts to, i. e. $\mathbb{P}^{\omega(do(i))}_X$.

Proof of Theorem 10 (Marginalisation of non-intervened variables). By Lemma 5 it suffices to proof this for marginalisation of one never-intervened-upon variable. Without loss of generality, let X_1 be the never-intervened-upon variable to be marginalised out. By acyclicity of the SEM \mathcal{M}_X , the structural equation corresponding to variable X_1 is of the form $X_1 = f_1(\mathbf{X}_{pa(1)}, E_1)$ and X_1 does not appear in the structural equation for any of its ancestors.

Now let $\mathcal{M}_Y = (S_Y, \mathcal{I}_Y, \mathcal{P}_F)$ be the SEM where

•
$$\mathcal{I}_Y = \mathcal{I}_X;$$

- $F_i = ((E_i, E_1) : i \in \mathbb{I}_X \setminus \{1\})$ are the noise variables distributed as implied by \mathbb{P}_E ;
- the structural equations S_Y are obtained from S_X by removing the structural equation of X_1 and replacing any occurrence of X_1 in the right-hand side of the structural equations of children of X_1 by $f_1(\mathbf{X}_{pa(1)}, E_1)$, yielding $X_i = f_i(f_1(\mathbf{X}_{pa(1)}, E_1), \mathbf{X}_{pa(i)}, E_i)$.

Note that the structural equations of the resulting SEM are still acyclic and are all of the form $X_i = h_i (\mathbf{X}_{\setminus i}, F_i)$. Then \mathcal{M}_Y is, by construction, an τ -exact transformation of \mathcal{M}_X for $\omega = \text{id}$.

C PROOF FOR SECTION 5.2: MICRO- TO MACRO-LEVEL

Proof of Theorem 11. We begin by defining a mapping between interventions

$$\omega : \mathcal{I}_X \to \mathcal{I}_Y$$

$$\emptyset \mapsto \emptyset$$

$$\operatorname{do}(W = w) \mapsto \operatorname{do}\left(\widehat{W} = \frac{1}{n}\sum_{i=1}^n w_i\right)$$

$$\operatorname{do}(Z = z) \mapsto \operatorname{do}\left(\widehat{Z} = \frac{1}{m}\sum_{i=1}^m z_i\right)$$

$$\operatorname{do}(W = w, Z = z) \mapsto \operatorname{do}\left(\widehat{W} = \frac{1}{n}\sum_{i=1}^n w_i, \widehat{Z} = \frac{1}{m}\sum_{i=1}^m z_i\right)$$

Note that ω is surjective and order-preserving (in fact, it is an order embedding). Therefore, it only remains to show that the distributions implied by $\tau(X)$ under any intervention $i \in \mathcal{I}_X$ agree with the corresponding distributions implied by \mathcal{M}_Y . That is, we have to show that

$$\mathbb{P}^{i}_{\tau(X)} = \mathbb{P}^{\operatorname{do}(\omega(i))}_{Y} \quad \forall i \in \mathcal{I}_{X}$$

In the observational setting, the distribution over \mathcal{Y} is implied by the following equations:

$$\widehat{W} = \frac{1}{n} \sum_{i=1}^{n} W_i = \frac{1}{n} \sum_{i=1}^{n} E_i$$
$$\widehat{Z} = \frac{1}{m} \sum_{i=1}^{m} Z_i = \frac{1}{m} \sum_{i=1}^{m} \left(\sum_{j=1}^{n} A_{ij} W_j + F_i \right) = \frac{a}{m} \widehat{W} + \frac{1}{m} \sum_{i=1}^{m} F_i$$

Since the distributions of the exogenous variables in \mathcal{M}_Y are given by $\hat{E} \sim \frac{1}{n} \sum_{i=1}^n E_i$, $\hat{F} \sim \frac{1}{m} \sum_{i=1}^m F_i$, it follows that $\mathbb{P}_{\tau(X)}^{\mathrm{do}(\emptyset)}$ and $\mathbb{P}_Y^{\mathrm{do}(\emptyset)}$ agree. Similarly, the push-forward measure on \mathcal{Y} induced by the intervention $\mathrm{do}(W = w) \in \mathcal{I}_X$ is given by

$$\widehat{W} = \frac{1}{n} \sum_{i=1}^{n} W_i = \frac{1}{n} \sum_{i=1}^{n} w_i$$
$$\widehat{Z} = \frac{1}{m} \sum_{i=1}^{m} Z_i = \frac{1}{m} \sum_{i=1}^{m} \left(\sum_{j=1}^{n} A_{ij} W_j + F_i \right) = \frac{a}{m} \widehat{W} + \frac{1}{m} \sum_{i=1}^{m} F_i$$

which is the same as the distribution induced by the ω -corresponding intervention do $\left(\widehat{W} = \frac{1}{n}\sum_{i=1}^{n}w_i\right)$ in \mathcal{M}_Y . Similar reasoning shows that this also holds for the interventions do(Z = z) and do(W = w, Z = z).

D PROOF FOR SECTION 5.3: STATIONARY BEHAVIOUR OF DYNAMICAL PROCESSES

Proof of Theorem 12. We begin by defining a mapping between interventions

$$\omega : \mathcal{I}_X \to \mathcal{I}_Y$$
$$\operatorname{do}(X_t^j = x_j \ \forall t \in \mathbb{Z}, \ \forall j \in J) \mapsto \operatorname{do}(Y^j = x_j \ \forall j \in J)$$

Note that ω is surjective and order-preserving (in fact, it is an order embedding). Therefore, it only remains to show that the distributions implied by $\tau(X)$ under any intervention $i \in \mathcal{I}_X$ agree with the corresponding distributions implied by \mathcal{M}_Y . That is, we have to show that

$$\mathbb{P}^{i}_{\tau(X)} = \mathbb{P}^{\operatorname{do}(\omega(i))}_{Y} \quad \forall i \in \mathcal{I}_{X}$$

For this we consider, without loss of generality, the distribution arising from performing the \mathcal{M}_X -level intervention

$$i = do(X_t^j = x_j \ \forall t \in \mathbb{Z}, \forall j \le m \le n) \in \mathcal{I}_X$$

for $m \in [n]$ (for m = 0 this amounts to the null-intervention).

Since A is a contraction mapping, it follows from Lemma 15 that for any intervention in \mathcal{I}_X , the sequence of random variables X_t defined by \mathcal{M}_X converges everywhere. That is, there exists a random variable X_* such that $X_t \xrightarrow[t \to \infty]{t \to \infty} X_*$. In the case of the intervention *i* above, the random variable X_* satisfies:

$$\begin{cases} X_*^k = x_k & \text{if } k \le m \\ X_*^k = \sum_j A_{kj} X_*^j + E^k & \text{if } m < k \le n \end{cases}$$
(1)

Since $\tau(X) = \lim_{t\to\infty} X_t$, it follows from the definition of X_* that $\tau(X) = X_*$, and hence $\tau(X)$ also satisfies the equations above. It follows (rewriting the second line in Equation 1 above) that under the push-forward measure $\mathbb{P}^i_{\tau(X)} = \tau\left(\mathbb{P}^{\mathrm{do}(i)}_X\right)$ the distribution of the random variable $\tau(X) = X_*$ is given by:

$$\begin{cases} X_*^k = x_k & \text{if } k \le m \\ X_*^k = \frac{\sum_{j \ne k} A_{kj} X_*^j}{1 - A_{kk}} + \frac{E^k}{1 - A_{kk}} & \text{if } m < k \le n \end{cases}$$

We need to compare this to the law of Y as implied by \mathcal{M}_Y under the intervention $\omega(i)$, i. e. $\mathbb{P}_Y^{\operatorname{do}(\omega(i))}$. The \mathcal{M}_Y -level intervention $\omega(i)$ corresponding to *i* is

$$\omega(i) = \operatorname{do}(Y^j = x_i \;\forall j \le m \le n) \in \mathcal{I}_Y$$

and so the structural equations of \mathcal{M}_{Y} under the intervention $\omega(do(i))$ are

$$\begin{cases} Y^k = x_k & \text{if } k \le m \\ Y^k = \frac{\sum_{j \ne k} A_{kj} Y^j}{1 - A_{kk}} + \frac{F^k}{1 - A_{kk}} & \text{if } m < k \le n \end{cases}$$

Since $F \sim E$ it indeed follows that $\tau(X) \sim Y$, i. e. $\mathbb{P}^{i}_{\tau(X)} = \mathbb{P}^{\operatorname{do}(\omega(i))}_{Y}$.

Thus \mathcal{M}_Y is an exact τ -transformation of \mathcal{M}_X .

D.1 CONTRACTION MAPPING AND CONVERGENCE

The following Lemmata show that A being a contraction mapping ensures that the sequence $(X_t)_{t \in \mathbb{Z}}$ defined by \mathcal{M}_X in Theorem 12 converges everywhere under any intervention $i \in \mathcal{I}_X$. That is, for any realisation $(x_t)_{t \in \mathbb{Z}}$ of this sequence, its limit $\lim_{t\to\infty} x_t$ as a sequence of elements of \mathbb{R}^n exists.

Lemma 13. Suppose that the function

$$f : \mathbb{R}^n \to \mathbb{R}^m$$
$$x \mapsto f(x)$$

is a contraction mapping. Then, for any $e \in \mathbb{R}^m$, so is the function

$$f^* : \mathbb{R}^n \to \mathbb{R}^m$$

 $x \mapsto f(x) + e$

Proof. By definition, there exists c < 1 such that for any $x, y \in \mathbb{R}^n$,

$$\|f^*(x) - f^*(y)\| = \|(f(x) + e) - (f(y) + e)\| = \|f(x) - f(y)\| \le c \|x - y\|$$

and hence f^* is a contraction mapping.

Lemma 14. Suppose that the function

$$f : \mathbb{R}^n \to \mathbb{R}^n$$
$$x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \mapsto \begin{pmatrix} f_1(x) \\ \vdots \\ f_n(x) \end{pmatrix}$$

is a contraction mapping. Then for any $m \leq n$, and $x_i^* \in \mathbb{R}$, $i \in [m]$, so is the function

$$f^* : \mathbb{R}^n \to \mathbb{R}^n$$
$$x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \mapsto \begin{pmatrix} x_1^* \\ \vdots \\ x_m^* \\ f_{m+1}(x) \\ \vdots \\ f_n(x) \end{pmatrix}$$

Proof. By definition, there exists c < 1 such that for any $x, y \in \mathbb{R}^n$,

$$\|f^{*}(x) - f^{*}(y)\| = \left\| \begin{pmatrix} x_{1}^{*} \\ \vdots \\ x_{m}^{*} \\ f_{m+1}(x) \\ \vdots \\ f_{n}(x) \end{pmatrix} - \begin{pmatrix} x_{1}^{*} \\ \vdots \\ x_{m}^{*} \\ f_{m+1}(y) \\ \vdots \\ f_{n}(y) \end{pmatrix} \right\| = \left\| \begin{pmatrix} 0 \\ \vdots \\ 0 \\ f_{m+1}(x) - f_{m+1}(y) \\ \vdots \\ f_{n}(x) - f_{n}(y) \end{pmatrix} \right\| \leq \left\| \begin{pmatrix} f_{1}(x) - f_{1}(y) \\ \vdots \\ f_{n}(x) - f_{n}(y) \end{pmatrix} \right\|$$
$$= \|f(x) - f(y)\|$$
$$\leq c \|x - y\|$$

and hence f^* is a contraction mapping.

Lemma 15. Consider the SEM \mathcal{M}_X in Theorem 12, and suppose that the linear map $A : \mathbb{R}^n \to \mathbb{R}^n$ is a contraction mapping. Then, for any intervention $i \in \mathcal{I}_X$, the sequence of X_t converges everywhere.

Proof. Consider, without loss of generality, the intervention

$$do(X_t^j = x_j \ \forall t \in \mathbb{Z}, \forall j \le m \le n) \in \mathcal{I}_X$$

for $m \in [n]$ (for m = 0 this amounts to the null-intervention). The structural equations under this intervention are

$$\begin{cases} X_{t+1}^k = x_k & \text{if } k \le m \\ X_{t+1}^k = \sum_j A_{kj} X_t^j + E^k & \text{if } m < k \le m \end{cases}$$

and thus the sequence X_t can be seen to transition according to the function $f = g \circ h$, where

$$h: \mathbb{R}^{n} \to \mathbb{R}^{n}$$
$$v \mapsto w = Av + E$$
$$g: \mathbb{R}^{n} \to \mathbb{R}^{n}$$
$$w = \begin{pmatrix} w_{1} \\ \vdots \\ w_{n} \end{pmatrix} \mapsto \begin{pmatrix} x_{1} \\ \vdots \\ x_{m} \\ w_{m+1} \\ \vdots \\ w_{n} \end{pmatrix}$$

By Lemma 13 and Lemma 14, f is a contraction mapping for any fixed E. Thus, by the contraction mapping theorem, the sequence of X_t converges everywhere to a unique fixed point.