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Feliu, Elisenda

Published in:
Proceedings of the Royal Society of Edinburgh Section A: Mathematics

DOI:
10.1098/rspa.2014.0530

Publication date:
2015

Document version
Peer reviewed version

Citation for published version (APA):
Feliu, E. (2015). Injectivity, multiple zeros, and multistationarity in reaction networks. Proceedings of the Royal Society of Edinburgh Section A: Mathematics, 471, [2173]. https://doi.org/10.1098/rspa.2014.0530

# Injectivity, multiple zeros, and multistationarity in reaction networks 

Elisenda Feliu

July 14, 2014


#### Abstract

Polynomial dynamical systems are widely used to model and study real phenomena. In biochemistry, they are the preferred choice for modeling the concentration of chemical species in reaction networks with mass-action kinetics. These systems are typically parameterised by many (unknown) parameters. A goal is to understand how properties of the dynamical systems depend on the parameters.

Qualitative properties relating to the behaviour of a dynamical system are locally inferred from the system at steady state. Here we focus on steady states that are the positive solutions to a parameterised system of generalised polynomial equations. In recent years, methods from computational algebra have been developed to understand these solutions, but our knowledge is limited: for example, we cannot efficiently decide how many positive solutions the system has as a function of the parameters. Even deciding whether there is one or more solutions is non-trivial.

We present a new method, based on so-called injectivity, to preclude or assert that multiple positive solutions exist. The results apply to generalised polynomials and variables can be restricted to the linear, parameter-independent first integrals of the dynamical system. The method has been tested in a wide range of systems.


## 1 Introduction

The cell's ability to respond on-off to gradual changes in a signal is linked to bistability, the ability of a dynamical system to admit two different stable steady states [1,2]. Specifically, bistability underlies the emergence of hysteresis and switch-like behavior.

In the context of molecular biology, models usually depend on unknown parameters and one is interested in determining whether there exists a choice of parameter values for which the system has two stable steady states. In particular, the concentration of the species in a reaction network is typically modelled with a system of ordinary differential equations (ODEs):

$$
\frac{d x}{d t}=f_{\kappa}(x), \quad x \in \mathbb{R}_{\geq 0}^{n},
$$

where $\kappa$ is a vector of parameters. Bistability arises when the equation $f_{\kappa}(x)=0$ admits two solutions that correspond to stable steady states. For unknown parameters, determining whether a steady state is stable is highly non-trivial. Therefore, the focus has mainly been on determining whether the equation $f_{\kappa}(x)=0$ admits multiple solutions for some parameter value $\kappa$.

[^0]Numerous methods have been developed to this end, e.g. [3-8]. These methods apply under different assumptions and to different modelling strategies. Most methods generally specialise to the steady states with positive coordinates, that is, they decide whether the function $f$ admits multiple positive zeros and exclude the zeros at the boundary. In this paper, we focus on injectivity-based methods, which underly several earlier approaches, for instance [5, $6,9,13]$. For polynomials, the setting was introduced in [9], and subsequently extended in [6, 14-17].

The idea underlying injectivity methods is simple: Consider a map $f: \mathbb{R}_{>0}^{n} \rightarrow \mathbb{R}^{n}$. If $f$ is injective, then there do not exist distinct $x, y$ in $\mathbb{R}_{>0}^{n}$ such that $f(x)=f(y)$. In particular, $f$ does not have two distinct zeros in $\mathbb{R}_{>0}^{n}$.

The converse is false. For instance, consider the map $f: \mathbb{R}_{>0}^{2} \rightarrow \mathbb{R}^{2}$ given by

$$
\begin{aligned}
& f_{1}\left(x_{1}, x_{2}\right)=x_{1} x_{2}-10 x_{1}+0.1 x_{2} \\
& f_{2}\left(x_{1}, x_{2}\right)=x_{1} x_{2}+0.1 x_{1}-10 x_{2}
\end{aligned}
$$

The only positive zero of $f$ is $(9.9,9.9)$, but $f$ is not injective on $\mathbb{R}_{>0}^{2}$ since

$$
f(5.9,2)=f(7.9,4)=(-47,-7.61)
$$

Thus, in this case, we cannot conclude that $f$ does not have multiple positive zeros by checking whether $f$ is injective. However, we can make another simple observation: The positive zeros of $f$ are also the positive solutions to:

$$
\begin{equation*}
x_{1}-10 x_{1} x_{2}^{-1}=-0.1, \quad x_{2}-10 x_{1}^{-1} x_{2}=-0.1 \tag{1}
\end{equation*}
$$

An easy check shows that the map $g: \mathbb{R}_{>0}^{2} \rightarrow \mathbb{R}^{2}$ defined by

$$
\begin{equation*}
g\left(x_{1}, x_{2}\right)=\left(x_{1}-10 x_{1} x_{2}^{-1}, x_{2}-10 x_{1}^{-1} x_{2}\right) \tag{2}
\end{equation*}
$$

is injective on $\mathbb{R}_{>0}^{2}$ and hence we conclude that neither (1) nor $f=0$ have multiple positive solutions. Observe that positivity of the solutions is critical here, because $g$ is obtained by dividing its two components by $x_{2}$ and $x_{1}$, respectively.

This simple example illustrates the approach introduced here: By checking the injectivity of a new map $g$ constructed from $f$, we conclude that $f$ does not have more than one positive zero. We focus on the positive zeros of so-called generalised polynomial maps, that is, polynomials with real exponents. For example, $g$ in (2) is a generalised polynomial map. More generally, we consider families of parameterised generalised polynomial maps (cf. (6) ) and use the idea above to preclude or determine the existence of multiple zeros on the positive part of affine subspaces $x^{*}+S$, where $x^{*} \in \mathbb{R}_{>0}^{n}$ and $S \subseteq \mathbb{R}^{n}$ is a vector subspace.

The method is designed to address multistationarity in reaction networks (see Section 3). However, the method focuses generally on the positive zeros of generalised polynomial maps. Therefore, we start by presenting the method in a mathematical framework in Section 2, In particular we explain how injectivity can be used both to preclude and to assert the existence of multiple positive zeros. We proceed to exemplify how the method can be used to address multistationarity for a wide range of reaction networks in Section 3. The steps of the method are summarised in subsection 22.4.

Given a positive integer $n$, we let $[n]:=\{1, \ldots, n\}$.

## 2 Mathematical framework

### 2.1 Injectivity and multiple zeros

S-injectivity. Given a vector subspace $S \subseteq \mathbb{R}^{n}$, we first define the notion of injectivity on the positive part of the affine subspaces $x^{*}+S$ with $x^{*} \in \mathbb{R}_{>0}^{n}$, cf. [16].

Definition 2.1. Let $f: \mathbb{R}_{>0}^{n} \rightarrow \mathbb{R}^{m}$ be a map and $S \subseteq \mathbb{R}^{n}$ a vector subspace. We say that

- $f$ is $S$-injective if for all distinct $x, y \in \mathbb{R}_{>0}^{n}$ such that $x-y \in S$ we have $f(x) \neq f(y)$.
- $f$ has multiple $S$-zeros if there exist distinct $x, y \in \mathbb{R}_{>0}^{n}$ such that $x-y \in S$ and $f(x)=f(y)=0$.

Equivalently, $f$ does not have multiple $S$-zeros if $f$ does not have two distinct zeros in some set $\left(x^{*}+S\right) \cap \mathbb{R}_{>0}^{n}$ with $x^{*} \in \mathbb{R}_{>0}^{n}$. Clearly, if $f$ is $S$-injective, then $f$ does not have multiple $S$-zeros.

Definition 2.2. Let $S \subseteq \mathbb{R}^{n}$ be a vector subspace and $\mathcal{F} \subseteq\left\{f: \mathbb{R}_{>0}^{n} \rightarrow \mathbb{R}^{m}\right\}$ be a set of functions. We say that

- $\mathcal{F}$ is $S$-injective, if $f$ is $S$-injective for all $f \in \mathcal{F}$.
- $\mathcal{F}$ has multiple $S$-zeros, if there exists $f \in \mathcal{F}$ such that $f$ has multiple $S$-zeros.

Generalised polynomial maps. We consider families of generalised polynomial maps given by two real matrices: $A=\left(a_{i j}\right) \in \mathbb{R}^{m \times r}$ (coefficient matrix) and $V=\left(v_{i j}\right) \in \mathbb{R}^{n \times r}$ (exponent matrix). We denote the $j$-th column of $V$ by $v^{j}$ and define, for $x \in \mathbb{R}_{>0}^{n}$, the vector $x^{V} \in \mathbb{R}_{>0}^{r}$ as

$$
\left(x^{V}\right)_{j}=x^{v^{j}}=x_{1}^{v_{1 j}} \cdot \ldots \cdot x_{n}^{v_{n j}}, \quad j \in[r] .
$$

The vector $x^{V}$ is a vector of generalised monomials, whose exponents are given by the columns of $V$. For a positive vector $\kappa \in \mathbb{R}_{>0}^{r}$, we define the map $f_{\kappa}: \mathbb{R}_{>0}^{n} \rightarrow \mathbb{R}^{m}$ as

$$
\begin{equation*}
f_{\kappa}(x)=A\left(\kappa \circ x^{V}\right), \tag{3}
\end{equation*}
$$

where $\circ$ denotes the product componentwise (or Hadamard product). The coefficient $\kappa_{j}$ of each monomial will be treated as a parameter. Since

$$
f_{\kappa}(x)=f_{\kappa}(y) \quad \text { if and only if } \quad A\left(\kappa \circ\left(x^{V}-y^{V}\right)\right)=0,
$$

then $S$-injectivity of $f_{\kappa}$ is unchanged if the matrix $A$ is replaced by a matrix $\widetilde{A}$ with the same kernel as $\widetilde{A}$.

For example, the map

$$
\begin{equation*}
f_{\kappa}(x)=\binom{\kappa_{1} x_{1} x_{2}-\kappa_{2} x_{2} x_{3}+\kappa_{3} x_{1} x_{4}}{\kappa_{2} x_{2} x_{3}-\kappa_{3} x_{1} x_{4}-\kappa_{4} x_{2} x_{4}}, \tag{4}
\end{equation*}
$$

takes the form (3) with

$$
A=\left(\begin{array}{rrrr}
1 & -1 & 1 & 0  \tag{5}\\
0 & 1 & -1 & -1
\end{array}\right), \quad V=\left(\begin{array}{rrrr}
1 & 0 & 1 & 0 \\
1 & 1 & 0 & 1 \\
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 1
\end{array}\right)
$$

such that $\kappa \circ x^{V}=\left(\kappa_{1} x_{1} x_{2}, \kappa_{2} x_{2} x_{3}, \kappa_{3} x_{1} x_{4}, \kappa_{4} x_{2} x_{4}\right)^{t}$.
Any generalised polynomial map can be written in the form (3)), possibly for more than one choice of $A, V$, and $\kappa$. For instance, a generalised polynomial map $f: \mathbb{R}_{>0}^{n} \rightarrow \mathbb{R}^{m}$ involving monomials $x^{v^{1}}, \ldots, x^{v^{r}}$ can be written componentwise as

$$
f_{i}(x)=\sum_{j=1}^{r} a_{i j} x^{v^{j}}, \quad a_{i j} \in \mathbb{R}, \quad i \in[n] .
$$

Hence, $f$ is of the form (3) with $\kappa=(1, \ldots, 1)^{t}, A=\left(a_{i j}\right)$ and $V$ having columns $v^{1}, \ldots, v^{r}$.
We let

$$
\begin{equation*}
\mathcal{F}_{A, V}:=\left\{f_{\kappa}: \mathbb{R}_{>0}^{n} \rightarrow \mathbb{R}^{m} \mid f_{\kappa}(x)=A\left(\kappa \circ x^{V}\right), \kappa \in \mathbb{R}_{>0}^{r}\right\} \tag{6}
\end{equation*}
$$

be the family of all generalised polynomial maps (3) obtained by varying the vector of parameters $\kappa$. Here, we address the question of determining whether $\mathcal{F}_{A, V}$ has multiple $S$-zeros. If the family $\mathcal{F}_{A, V}$ is $S$-injective, then $\mathcal{F}_{A, V}$ does not have multiple $S$-zeros, but the converse is not true.

In [16], a characterisation of the families of generalised polynomial maps $\mathcal{F}_{A, V}$ that are $S$-injective is given in terms of sign vectors and symbolic determinants. Namely, the family is $S$-injective if and only if some particular sets of signs do not intersect. These sets are constructed from the orthants of $\mathbb{R}^{n}$ and $\mathbb{R}^{r}$ that certain subspaces, defined from $A, V, S$, intersect.

When $\operatorname{dim}(S) \geq \operatorname{rank}(A), S$-injectivity of $\mathcal{F}_{A, V}$ can also be decided based on the signs of the coefficients of the determinant of a symbolic matrix. The latter scenario is applicable when studying steady states of chemical reaction networks. Due to its simplicity and applicability, we describe only the determinant-based criterion for injectivity here. The reader is referred to [16] for the sign criterion.

Determinant criterion for injectivity Let $S \subseteq \mathbb{R}^{n}$ be a vector subspace of dimension $s$ and $A \in \mathbb{R}^{m \times r}$ such that $s \geq \operatorname{rank}(A)$. Let $Z \in \mathbb{R}^{(n-s) \times n}$ be any matrix whose rows are a basis of $S^{\perp}$ and choose $\widetilde{A} \in \mathbb{R}^{s \times r}$ such that $\operatorname{ker}(A)=\operatorname{ker}(\widetilde{A})$. For example, we can choose a set of $s$ rows of $A$ with the same rank as $A$.

For $\kappa \in \mathbb{R}_{>0}^{r}$ and $\lambda \in \mathbb{R}_{>0}^{n}$, let $M_{\kappa, \lambda}$ be the square matrix given in block form as

$$
\begin{equation*}
M_{\kappa, \lambda}=\binom{Z}{\widetilde{A} \operatorname{diag}(\kappa) V^{t} \operatorname{diag}(\lambda)}, \tag{7}
\end{equation*}
$$

where, for a vector $w, \operatorname{diag}(w)$ denotes the diagonal matrix with diagonal $w$. By considering $\kappa, \lambda$ as indeterminates, the determinant of $M_{\kappa, \lambda}$ is a polynomial $p(\kappa, \lambda)$ in $\kappa, \lambda$. It is a result of [6, 16], that the family $\mathcal{F}_{A, V}$ is $S$-injective if and only if $p(\kappa, \lambda)$ is not identically zero and, further, all its coefficients have the same sign.

Note than when $\operatorname{rank}(A)<s$, then $p(\kappa, \lambda)$ is identically zero and the family $\mathcal{F}_{A, V}$ is not $S$-injective.

### 2.2 Gauss reduction and injectivity

In this subsection we present the steps of the method that can be used to conclude that $\mathcal{F}_{A, V}$ does not have multiple $S$-zeros.

The family $\mathcal{G}_{A, V}$. Let $S \subseteq \mathbb{R}^{n}$ be a vector subspace of dimension $s$ and assume that $A$ has rank $s$. We choose as above a matrix $\widetilde{A} \in \mathbb{R}^{s \times r}$ such that $\operatorname{ker}(A)=\operatorname{ker}(\widetilde{A})$ and let $B$ be the Gauss reduction of $\widetilde{A}$. Since the kernel of $A$ and $B$ agree, the family $\mathcal{F}_{B, V}$ is $S$-injective or has multiple $S$-zeros if and only if this is the case for $\mathcal{F}_{A, V}$. By reordering the columns if necessary, we can assume that columns 1 to $s$ of $\widetilde{A}$ are linearly independent. Therefore, without loss of generality, we restrict to families $\mathcal{F}_{A, V}$ with

$$
\begin{equation*}
A=\left(i d_{s} \mid A_{1}\right) \tag{8}
\end{equation*}
$$

where $i d_{s}$ is the identity matrix of size $s$ and $A_{1} \in \mathbb{R}^{s \times(r-s)}$. Then $f_{\kappa} \in \mathcal{F}_{A, V}$ satisfies $f_{\kappa}(x)=0$ if and only if

$$
\left(\begin{array}{c}
\kappa_{1} x^{v^{1}} \\
\vdots \\
\kappa_{s} x^{v^{s}}
\end{array}\right)=-A_{1}\left(\begin{array}{c}
\kappa_{s+1} x^{v^{s+1}} \\
\vdots \\
\kappa_{r} x^{v^{r}}
\end{array}\right)
$$

and this equality holds for $x \in \mathbb{R}_{>0}^{n}$ if and only if

$$
\left(\begin{array}{c}
\kappa_{1}  \tag{9}\\
\vdots \\
\kappa_{s}
\end{array}\right)=g_{\widehat{\kappa}}(x), \quad \text { where } \quad g_{\widehat{\kappa}}(x)=-\left(\begin{array}{c}
x^{-v^{1}} \\
\vdots \\
x^{-v^{s}}
\end{array}\right) \circ A_{1}\left(\begin{array}{c}
\kappa_{s+1} x^{v^{s+1}} \\
\vdots \\
\kappa_{r} x^{v^{r}}
\end{array}\right)
$$

with $\widehat{\kappa}=\left(\kappa_{s+1}, \ldots, \kappa_{r}\right)$. If the map $g_{\widehat{\kappa}}(x)$ is $S$-injective, then $f_{\kappa}$ does not have multiple $S$-zeros. Therefore, by defining

$$
\mathcal{G}_{A, V}:=\left\{g_{\widehat{\kappa}}: \mathbb{R}_{>0}^{n} \rightarrow \mathbb{R}^{s} \mid \widehat{\kappa} \in \mathbb{R}_{>0}^{r-s}\right\}
$$

we have that $\mathcal{F}_{A, V}$ does not have multiple $S$-zeros if $\mathcal{G}_{A, V}$ is $S$-injective.

Example 1. Consider the family $\mathcal{F}_{A, V}$ defined by (5) and let

$$
\begin{equation*}
S=\langle(1,-1,0,0),(0,0,1,-1)\rangle \subseteq \mathbb{R}^{4} \tag{10}
\end{equation*}
$$

Both the dimension of $S$ and the rank of $A$ are 2. Hence we take $\widetilde{A}=A$ and replace $A$ by its Gauss reduction:

$$
A=\left(\begin{array}{rrrr}
1 & 0 & 0 & -1  \tag{11}\\
0 & 1 & -1 & -1
\end{array}\right)
$$

The determinant criterion tells us that the family $\mathcal{F}_{A, V}$ is not $S$-injective. The map $g_{\widehat{\kappa}}$ in (9) is

$$
\begin{align*}
g_{\kappa_{3}, \kappa_{4}}(x) & =-\binom{x_{1}^{-1} x_{2}^{-1}}{x_{2}^{-1} x_{3}^{-1}} \circ\left(\begin{array}{rr}
0 & -1 \\
-1 & -1
\end{array}\right)\binom{\kappa_{3} x_{1} x_{4}}{\kappa_{4} x_{2} x_{4}}=\binom{\kappa_{4} x_{1}^{-1} x_{4}}{\kappa_{3} x_{1} x_{2}^{-1} x_{3}^{-1} x_{4}+\kappa_{4} x_{3}^{-1} x_{4}} \\
& =\left(\begin{array}{lll}
1 & 0 & 0 \\
0 & 1 & 1
\end{array}\right)\left(\begin{array}{c}
\kappa_{4} x_{1}^{-1} x_{4} \\
\kappa_{3} x_{1} x_{2}^{-1} x_{3}^{-1} x_{4} \\
\kappa_{4} x_{3}^{-1} x_{4}
\end{array}\right) . \tag{12}
\end{align*}
$$

The family $\mathcal{G}_{A, V}$ is not of the form (6), since two different monomials have the same parameter as coefficient. However, observe that

$$
\mathcal{G}_{A, V}=\left\{f_{\eta} \in \mathcal{F}_{A^{\prime}, V^{\prime}} \mid \eta \in \mathbb{R}_{>0}^{3}, \quad \eta_{1}=\eta_{3}\right\} \subseteq \mathcal{F}_{A^{\prime}, V^{\prime}}
$$

with

$$
A^{\prime}=\left(\begin{array}{lll}
1 & 0 & 0  \tag{13}\\
0 & 1 & 1
\end{array}\right), \quad V^{\prime}=\left(\begin{array}{rrr}
-1 & 1 & 0 \\
0 & -1 & 0 \\
0 & -1 & -1 \\
1 & 1 & 1
\end{array}\right)
$$

We use the determinant criterion to decide whether the family $\mathcal{F}_{A^{\prime}, V^{\prime}}$ is $S$-injective. For suitable $Z$, the matrix $M_{\kappa, \lambda}$ is

$$
M_{\kappa, \lambda}=\left(\begin{array}{cccc}
1 & 1 & 0 & 0 \\
0 & 0 & 1 & 1 \\
-\kappa_{1} \lambda_{1} & 0 & 0 & \kappa_{1} \lambda_{4} \\
\kappa_{2} \lambda_{1} & -\kappa_{2} \lambda_{2} & -\left(\kappa_{2}+\kappa_{3}\right) \lambda_{3} & \left(\kappa_{2}+\kappa_{3}\right) \lambda_{4}
\end{array}\right)
$$

The determinant of this matrix is

$$
\operatorname{det}\left(M_{\kappa, \lambda}\right)=-\kappa_{1}\left(\kappa_{2} \lambda_{1} \lambda_{3}+2 \kappa_{2} \lambda_{1} \lambda_{4}+\kappa_{2} \lambda_{2} \lambda_{4}+\kappa_{3} \lambda_{1} \lambda_{3}+\kappa_{3} \lambda_{1} \lambda_{4}\right)
$$

Since all coefficients have the same sign, the family $\mathcal{F}_{A^{\prime}, V^{\prime}}$ is $S$-injective. Hence, so is the family $\mathcal{G}_{A, V}$ and it follows that $\mathcal{F}_{A, V}$ does not have multiple $S$-zeros.

The family $\mathcal{F}_{A^{\prime}, V^{\prime}}$. Although the family $\mathcal{G}_{A, V}$ might not be of the type (6), one can always find a family $\mathcal{F}_{A^{\prime}, V^{\prime}}$, with $A^{\prime} \in \mathbb{R}^{s \times r^{\prime}}, V^{\prime} \in \mathbb{R}^{n \times r^{\prime}}$, and such that

$$
\mathcal{G}_{A, V} \subseteq \mathcal{F}_{A^{\prime}, V^{\prime}} .
$$

Indeed, one first identifies the different generalised monomials in $\widehat{\kappa}$, and $x$ of $g_{\widehat{\kappa}}(x)$. The exponents of the monomials in $x$ define $V^{\prime}$. The matrix $A^{\prime}$ is the matrix of coefficients. When doing so, some parameter $\kappa_{j}$ might be multiplying two different generalised monomials in $x$, say $x^{w_{1}}$ and $x^{w_{2}}$, as it is the case in the example. This occurs whenever a column $j$ of $A_{1}$ contains two nonzero entries at rows $i_{1}, i_{2}$ such that $v^{i_{1}} \neq v^{i_{2}}$. Then the monomial $\kappa_{s+j} x^{v^{s+j}}$ appears as a summand in two different rows of the product

$$
A_{1}\left(\kappa \circ\left(x^{v^{s+1}}, \ldots, x^{v^{r}}\right)\right)^{t}
$$

and these rows are multiplied by different monomials $x^{-v_{1}}, x^{-v^{i_{2}}}$ when constructing $g_{\widehat{\kappa}}$. In general, there exists a partition $\left[r^{\prime}\right]=I_{1} \cup \cdots \cup I_{q}$ and matrices $A^{\prime} \in \mathbb{R}^{s \times r^{\prime}}$ and $V^{\prime} \in \mathbb{R}^{n \times r^{\prime}}$ such that

$$
\begin{equation*}
\mathcal{G}_{A, V}=\left\{f_{\eta} \in \mathcal{F}_{A^{\prime}, V^{\prime}} \mid \eta \in \mathbb{R}_{>0}^{r^{\prime}}, \eta_{i}=\eta_{j} \text { if } i, j \in I_{k}, \text { for some } k \in[q]\right\} \tag{14}
\end{equation*}
$$

In the example above, $r^{\prime}=3$ and we let $I_{1}=\{2\}, I_{2}=\{1,3\}$.

Summary of the steps. Given a family of generalised polynomial maps $\mathcal{F}_{A, V}$ such that $A$ is Gauss reduced and of the form (8), then we construct the families $\mathcal{G}_{A, V}$ and $\mathcal{F}_{A^{\prime}, V^{\prime}}$. Since the matrix $A^{\prime}$ has $s$ rows, we apply the determinant criterion to determine whether $\mathcal{F}_{A^{\prime}, V^{\prime}}$ is $S$-injective. If this is the case, then we conclude that $\mathcal{F}_{A, V}$ does not have multiple $S$-zeros.

Observe that when constructing $\mathcal{G}_{A, V}$, the parameters $\kappa_{1}, \ldots, \kappa_{s}$ are selected and "eliminated". We could have selected any other set of $s$ parameters $\kappa_{i_{1}}, \ldots, \kappa_{i_{s}}$, as long as the columns $i_{1}, \ldots, i_{s}$ of $A$ are linearly independent. In this case, the parameters $\kappa_{i_{1}}, \ldots, \kappa_{i_{s}}$ are eliminated if we first permute the columns of $A$ and $V$, and the entries of $\kappa$ simultaneously such
that the indices $i_{1}, \ldots, i_{s}$ are the first indices. We then apply Gauss reduction and proceed as above with the new data.

Therefore, if $\mathcal{F}_{A^{\prime}, V^{\prime}}$ is not $S$-injective, then we repeat the described process after simultaneously permuting the columns of $A$ and $V$, and the entries of $\kappa$. We do this for all possible permutations such that the first $s$ columns of $A$ are linearly independent, unless for some permutation we conclude that $\mathcal{F}_{A, V}$ does not have multiple $S$-zeros and the procedure stops.

The family $\mathcal{F}_{\widehat{A}, \widehat{V}}$. In preparation for the next section, in particular for the proof of Proposition 2.1, we embed $\mathcal{G}_{A, V}$ into an even larger family $\mathcal{F}_{\widehat{A}, \widehat{V}}$ where:

- The matrix $\widehat{A} \in \mathbb{R}^{s \times s(r-s)}$ is the block diagonal matrix with diagonal blocks given by minus the rows of $A_{1}=\left(\bar{a}_{i j}\right)$ :

$$
\widehat{a}_{i,(i-1)(r-s)+j}=-\bar{a}_{i j}, \quad \text { for } \quad i \in[s], j \in[r-s],
$$

and zero otherwise.

- The $\ell$-th column of $\widehat{V} \in \mathbb{R}^{n \times s(r-s)}$ is

$$
v^{j+s}-v^{i}, \quad \text { if } \quad \ell=(i-1)(r-s)+j, \quad \text { with } \quad i \in[s], j \in[r-s] .
$$

Given $\varphi=f_{\eta} \in \mathcal{F}_{\widehat{A}, \widehat{V}}$ with $\eta \in \mathbb{R}_{>0}^{s(r-s)}$, then $\varphi \in \mathcal{G}_{A, V}$ if and only if

$$
\eta_{j+(b-1)(r-s)}=\eta_{j+(c-1)(r-s)}
$$

for all $j \in[r-s]$ and $b, c \in[s]$ such that $\bar{a}_{b j} \bar{a}_{c j} \neq 0$.
For each $j \in[r-s]$, let $\alpha_{j} \subseteq\{1, \ldots, s\}$ be the support of the $j$-th column of $A_{1}$. Then,

$$
\begin{equation*}
\mathcal{G}_{A, V}=\left\{f_{\eta} \in \mathcal{F}_{\widehat{A}, \hat{V}} \mid \eta_{j+(b-1)(r-s)}=\eta_{j+(c-1)(r-s)}, \text { if } b, c \in \alpha_{j}\right\} \tag{15}
\end{equation*}
$$

Observe that only the sets $\alpha_{j}$ with cardinality at least 2 are relevant. Further, we have

$$
\mathcal{G}_{A, V} \subseteq \mathcal{F}_{A^{\prime}, V^{\prime}} \subseteq \mathcal{F}_{\widehat{A}, \widehat{V}}
$$

For example, consider the matrix $V$ in (5) and $A$ in (11), where $s=2, r=4$, and $n=4$. We have

$$
-A_{1}=\left(\begin{array}{rr}
0 & 1 \\
1 & 1
\end{array}\right), \quad \widehat{A}=\left(\begin{array}{rrrr}
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 1
\end{array}\right), \quad \text { and } \quad \widehat{V}=\left(\begin{array}{rrrr}
0 & -1 & 1 & 0 \\
-1 & 0 & -1 & 0 \\
0 & 0 & -1 & -1 \\
1 & 1 & 1 & 1
\end{array}\right) .
$$

A map $f_{\eta} \in \mathcal{F}_{\widetilde{A}, \widetilde{V}}$ is of the form

$$
\begin{aligned}
f_{\eta}(x) & =\widetilde{A}\left(\left(\eta_{1} x_{2}^{-1} x_{4}, \eta_{2} x_{1}^{-1} x_{4}, \eta_{3} x_{1} x_{2}^{-1} x_{3}^{-1} x_{4}, \eta_{4} x_{3}^{-1} x_{4}\right)^{t}\right) \\
& =\left(\eta_{2} x_{1}^{-1} x_{4}, \eta_{3} x_{1} x_{2}^{-1} x_{3}^{-1} x_{4}+\eta_{4} x_{3}^{-1} x_{4}\right)^{t}
\end{aligned}
$$

The supports of the columns of $A_{1}$ are given by $\alpha_{1}=\{2\}, \alpha_{2}=\{1,2\}$. Then, the description of $\mathcal{G}_{A, V}$ according to (15) is

$$
\mathcal{G}_{A, V}=\left\{f_{\eta} \in \mathcal{F}_{\widehat{A}, \widehat{V}} \mid \eta_{2}=\eta_{4}\right\}
$$

which agrees with (12).

### 2.3 Gauss reduction and multiple zeros

There exist families $\mathcal{F}_{A, V}$ that do not have multiple $S$-zeros but for which the families $\mathcal{F}_{A^{\prime}, V^{\prime}}$ are not $S$-injective for all possible column permutations of $A$. Therefore, the steps outlined above do not guarantee the existence of multiple $S$-zeros for some member of $\mathcal{F}_{A, V}$. However, under some extra conditions, the existence of multiple $S$-zeros can be asserted. These conditions are described in this section.

Example 2. Let $A$ be as in (11) and let $V$ be

$$
V=\left(\begin{array}{rrrr}
1 & 0 & 1 & -1  \tag{16}\\
1 & 1 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 1
\end{array}\right)
$$

With this matrix $V$, the map $g_{\widehat{\kappa}}(x)$ in (12) becomes

$$
g_{\kappa_{3}, \kappa_{4}}(x)=\left(\begin{array}{ccc}
1 & 0 & 0  \tag{17}\\
0 & 1 & 1
\end{array}\right)\left(\begin{array}{c}
\kappa_{4} x_{1}^{-1} x_{4} \\
\kappa_{3} x_{1} x_{2}^{-1} x_{3}^{-1} x_{4} \\
\kappa_{4} x_{1}^{-1} x_{2}^{-1} x_{3}^{-1} x_{4}
\end{array}\right) .
$$

Then, $\mathcal{G}_{A, V}$ is included in $\mathcal{F}_{A^{\prime}, V^{\prime}}$ with $A^{\prime}$ and $V^{\prime}=\left(v_{i j}^{\prime}\right)$ given in (13), except for the last column of $V^{\prime}$ where $v_{13}^{\prime}=v_{23}^{\prime}=-1$. With $S$ given in (10), the family $\mathcal{F}_{A^{\prime}, V^{\prime}}$ is not $S$-injective. Indeed, for a suitable $Z$, the matrix $M_{\kappa, \lambda}$ is

$$
M_{\kappa, \lambda}=\left(\begin{array}{cccc}
1 & 1 & 0 & 0 \\
0 & 0 & 1 & 1 \\
\kappa_{1} \lambda_{1} & 0 & 0 & \kappa_{1} \lambda_{4} \\
\left(\kappa_{2}-\kappa_{3}\right) \lambda_{1} & -\left(\kappa_{2}+\kappa_{3}\right) \lambda_{2} & -\left(\kappa_{2}+\kappa_{3}\right) \lambda_{3} & \left(\kappa_{2}+\kappa_{3}\right) \lambda_{4}
\end{array}\right) .
$$

We have that

$$
\operatorname{det}\left(M_{\kappa, \lambda}\right)=\kappa_{1}\left(\kappa_{2} \lambda_{1} \lambda_{3}+2 \kappa_{3} \lambda_{1} \lambda_{4}-\kappa_{2} \lambda_{2} \lambda_{4}+\kappa_{3} \lambda_{1} \lambda_{3}-\kappa_{3} \lambda_{2} \lambda_{4}\right)
$$

which has both positive and negative coefficients. Therefore, there exist $\eta \in \mathbb{R}_{>0}^{3}$ and distinct $x, y \in \mathbb{R}_{>0}^{4}$ such that $x-y \in S$ and $A^{\prime}\left(\eta \circ x^{V^{\prime}}\right)=A^{\prime}\left(\eta \circ y^{V^{\prime}}\right)$. That is,

$$
\begin{aligned}
\eta_{1} x_{1}^{-1} x_{4} & =\eta_{1} y_{1}^{-1} y_{4} \\
\eta_{2} x_{1} x_{2}^{-1} x_{3}^{-1} x_{4}+\eta_{3} x_{1}^{-1} x_{2}^{-1} x_{3}^{-1} x_{4} & =\eta_{2} y_{1} y_{2}^{-1} y_{3}^{-1} y_{4}+\eta_{3} y_{1}^{-1} y_{2}^{-1} y_{3}^{-1} y_{4}
\end{aligned}
$$

The first equality holds independently of $\eta_{1}$. Hence for $\kappa_{3}=\eta_{2}$ and $\kappa_{4}=\eta_{3}$, we have $g_{\kappa_{3}, \kappa_{4}}(x)=g_{\kappa_{3}, \kappa_{4}}(y)$. We define $\kappa_{1}, \kappa_{2}>0$ as

$$
\binom{\kappa_{1}}{\kappa_{2}}=g_{\kappa_{3}, \kappa_{4}}(x) \quad\left(=g_{\kappa_{3}, \kappa_{4}}(y)\right) .
$$

Reversing the steps from $f_{\kappa}$ to $g_{\widehat{\kappa}}$, it follows that

$$
A\left(\kappa \circ x^{V}\right)=A\left(\kappa \circ y^{V}\right)=0, \quad x-y \in S, \quad x \neq y
$$

and thus $\mathcal{F}_{A, V}$ has multiple $S$-zeros.

Asserting multiple $S$-zeros. Assume that there exist parameters $\widehat{\kappa} \in \mathbb{R}_{>0}^{r-s}$ and distinct $x, y \in \mathbb{R}_{>0}^{n}$ with $x-y \in S$ such that

$$
\begin{align*}
& g_{\widehat{\kappa}}(x)=g_{\widehat{\kappa}}(y),  \tag{18}\\
& g_{\widehat{\kappa}}(x) \in \mathbb{R}_{>0}^{s} \tag{19}
\end{align*}
$$

Define $\kappa \in \mathbb{R}_{>0}^{r}$ by $\left(\kappa_{1}, \ldots, \kappa_{s}\right)=g_{\widehat{\kappa}}(x)$ and $\left(\kappa_{s+1}, \ldots, \kappa_{r}\right)=\widehat{\kappa}$. Then we have that $A\left(\kappa \circ x^{V}\right)=$ $A\left(\kappa \circ y^{V}\right)=0$, and hence the family $\mathcal{F}_{A, V}$ has multiple $S$-zeros.

To check whether there exist $\widehat{\kappa}, x, y$ such that (18) and (19) are fulfilled, we follow the following strategy.

Checking (18). Assume that $\mathcal{F}_{A^{\prime}, V^{\prime}}$ is not $S$-injective. Then neither is $\mathcal{F}_{\widehat{A}, \widehat{V}}$ and there exist parameters $\eta \in \mathbb{R}_{>0}^{s(r-s)}$ and distinct $x, y \in \mathbb{R}_{>0}^{n}$ with $x-y \in S$ such that $f_{\eta}(x)=f_{\eta}(y)$, for $f_{\eta} \in \mathcal{F}_{\widehat{A}, \widehat{V}}$. We can manually inspect the form of such a vector $\eta$ and obtain one such that $f_{\eta}$ belongs to $G_{A, V}$.

A sufficient and implementable condition for (18) is given by the next proposition. The idea is that in order to conclude that (18) holds for some $\widehat{\kappa} \in \mathbb{R}_{>0}^{r-s}$, we need sufficient freedom to modify $\eta$ to satisfy the relations in (15).

By modify we mean the following. Given $\epsilon \in \mathbb{R}_{>0}^{s}$, define $\widehat{\epsilon} \in \mathbb{R}_{>0}^{s(r-s)}$ by

$$
\widehat{\epsilon}_{\ell}=\epsilon_{i}, \quad \text { if } \quad \ell=j+(i-1)(r-s), \quad \text { with } i \in[s], j \in[r-s]
$$

that is, $\widehat{\epsilon}=\left(\epsilon_{1}, \ldots, \epsilon_{1}, \ldots, \epsilon_{s}, \ldots, \epsilon_{s}\right)$ such that each $\epsilon_{i}$ is repeated $r-s$ times.
Let $\widehat{\eta}=\widehat{\epsilon} \circ \eta \in \mathbb{R}_{>0}^{s(r-s)}$. If $f_{\eta} \in \mathcal{F}_{\widehat{A}, \widehat{V}}$, then $\epsilon f_{\eta}=f_{\widehat{\eta}} \in \mathcal{F}_{\widehat{A}, \widehat{V}}$. Therefore, if $\widehat{\eta}$ fulfils the conditions in (15), then $f_{\widehat{\eta}} \in \mathcal{G}_{A, V}, f_{\widehat{\eta}}(x)=f_{\widehat{\eta}}(y)$, and (18) is fulfilled.

Proposition 2.1. Let $\eta \in \mathbb{R}_{>0}^{s(r-s)}$. Assume that for each $j \in[r-s]$ there exists $\ell_{j} \in[s]$ such that $\alpha_{j} \cap \alpha_{j^{\prime}}=\left\{\ell_{j}\right\}=\left\{\ell_{j^{\prime}}\right\}$ for all $j, j^{\prime}$ such that the cardinality of $\alpha_{j}, \alpha_{j^{\prime}}$ is at least two and such that $\alpha_{j} \cap \alpha_{j^{\prime}} \neq \emptyset$.

Then there exists $\epsilon \in \mathbb{R}_{>0}^{s}$ such that the vector $\widehat{\epsilon} \circ \eta \in \mathbb{R}_{>0}^{s(r-s)}$ fulfils the conditions in (15).
Proof. Let $b \in[s]$. If $b \in \bigcup_{\left\{j \mid \# \alpha_{j}>1\right\}} \alpha_{j}$, choose $j$ such that $b \in \alpha_{j}$ and define

$$
\epsilon_{b}=\frac{\eta_{j+\left(\ell_{j}-1\right)(r-s)}}{\eta_{j+(b-1)(r-s)}} .
$$

This is well defined because if $b \in \alpha_{j} \cap \alpha_{j^{\prime}}$ then $b=\ell_{j}=\ell_{j^{\prime}}$ and $\epsilon_{b}=1$. Define $\epsilon_{b}=1$ otherwise. Let $\widehat{\eta}=\widehat{\epsilon} \circ \eta$. If $b \in \alpha_{j}$ and $\# \alpha_{j}>1$, then

$$
\widehat{\eta}_{j+(b-1)(r-s)}=\frac{\eta_{j+\left(\ell_{j}-1\right)(r-s)}}{\eta_{j+(b-1)(r-s)}} \eta_{j+(b-1)(r-s)}=\eta_{j+\left(\ell_{j}-1\right)(r-s)}
$$

and hence the conditions in (15) are fulfilled.
Therefore, if $\mathcal{F}_{A^{\prime}, V^{\prime}}$ is not $S$-injective and the assumptions in Proposition 2.1 are fulfilled, then there exists $\widehat{\kappa} \in \mathbb{R}_{>0}^{r-s}$ such that (18) holds. The conditions of the statement of Proposition 2.1 are clearly fulfilled if the columns of $A_{1}$ have disjoint supports.

Checking (19). If the non-zero entries of $A_{1}$ are all negative and each row has at least one negative entry, then condition (19) is fulfilled for all $\widehat{\kappa}$. This is the case in the example above. If this is not the case, then we can resort to the following result, which is adapted from Lemma 4.1 in 9 .

Proposition 2.2. Let $A^{\prime} \in \mathbb{R}^{s \times r^{\prime}}, V^{\prime} \in \mathbb{R}^{n \times r^{\prime}}, S \subseteq \mathbb{R}^{n}$ a vector subspace, and $M_{\kappa, \lambda}$ as defined in (77) for some choice of $Z$. Let $\eta \in \mathbb{R}_{>0}^{r^{\prime}}$ such that

$$
A^{\prime} \eta \in \mathbb{R}_{>0}^{s} \quad \text { and } \quad \operatorname{det}\left(M_{\eta, \lambda}\right)=0, \quad \text { for some } \lambda \in \mathbb{R}_{>0}^{n}
$$

Then there exist distinct $x, y \in \mathbb{R}_{>0}^{n}$ and $\kappa \in \mathbb{R}_{>0}^{r^{\prime}}$ such that $x-y \in S$ and

$$
A^{\prime}\left(\kappa \circ x^{V^{\prime}}\right)=A^{\prime}\left(\kappa \circ y^{V^{\prime}}\right) \in \mathbb{R}_{>0}^{s}
$$

Proof. The matrix $A^{\prime} \operatorname{diag}(\eta) V^{\prime t} \operatorname{diag}(\lambda)$ has nontrivial kernel in $S$, since $\operatorname{det}\left(M_{\eta, \lambda}\right)=0$. That is, there exists $\gamma \in \operatorname{ker}\left(A^{\prime} \operatorname{diag}(\eta) V^{\prime t} \operatorname{diag}(\lambda)\right) \cap S, \gamma \neq 0$. We define $x, y \in \mathbb{R}_{>0}^{n}$ and $\kappa \in \mathbb{R}_{>0}^{r^{\prime}}$ by

$$
\begin{aligned}
x_{i} & = \begin{cases}\gamma_{i} /\left(e^{\gamma_{i} \lambda_{i}}-1\right) & \text { if } \gamma_{i} \neq 0 \\
1 & \text { otherwise }\end{cases} \\
y_{i} & =x_{i} e^{\gamma_{i} \lambda_{i}} \\
\kappa & =\left(\eta \circ V^{\prime t}(\lambda \circ \gamma)\right) /\left(y^{V^{\prime}}-x^{V^{\prime}}\right),
\end{aligned}
$$

where division is componentwise and $0 / 0=1$. It is easy to check that $y-x=\gamma \in S, \kappa \in \mathbb{R}_{>0}^{r^{\prime}}$, and $A^{\prime}\left(\kappa \circ\left(y^{V^{\prime}}-x^{V^{\prime}}\right)\right)=0$, cf. [4, Section 7], [15, Th. 5.6]. By replacing $\gamma$ by $\epsilon \gamma$ for $\epsilon>0$ in the definitions above, define analogously $x_{\epsilon}, y_{\epsilon}$. We have

$$
\begin{aligned}
A^{\prime}\left(\kappa \circ x_{\epsilon}^{V^{\prime}}\right) & =A^{\prime}\left(\frac{\eta \circ V^{\prime t}(\lambda \circ \epsilon \gamma) \circ x_{\epsilon}^{V^{\prime}}}{y_{\epsilon}^{V^{\prime}}-x_{\epsilon}^{V^{\prime}}}\right)=A^{\prime}\left(\frac{\eta \circ V^{\prime t}(\lambda \circ \epsilon \gamma)}{e^{V^{\prime t}(\lambda \circ \epsilon \gamma)}-1}\right) \\
& \xrightarrow{\epsilon \rightarrow 0} A^{\prime}\left(\eta \circ(1, \ldots, 1)^{t}\right)=A^{\prime} \eta \in \mathbb{R}_{>0}^{s}
\end{aligned}
$$

Therefore, for $\epsilon$ small enough, we get the desired result.

### 2.4 The method

The procedures described in subsections 22.2 and 22.3 give a new injectivity-based method to determine whether a set of functions defined by generalised polynomial maps in the positive orthant admit more than one zero on the positive part of $x^{*}+S$ for varying $x^{*}$.

Specifically, given $A, V, S$, we proceed as follows:
0. Check whether $\mathcal{F}_{A, V}$ is $S$-injective. If yes the family $\mathcal{F}_{A, V}$ does not admit multiple $S$-zeros and stop. If not, proceed.

1. Compute the Gauss reduction of $A$ and the function $g_{\widehat{\kappa}}$.
2. Identify matrices $A^{\prime}, V^{\prime}$ such that $g_{\widehat{\kappa}} \in \mathcal{F}_{A^{\prime}, V^{\prime}}$.
3. Check whether the family $\mathcal{F}_{A^{\prime}, V^{\prime}}$ is $S$-injective. If yes, the family $\mathcal{F}_{A, V}$ does not admit multiple $S$-zeros and stop. If not, proceed.
4. Check whether the assumptions in Proposition 2.1 are fulfilled. If not, go to step 7 . If yes, proceed.
5. If all nonzero entries of $A_{1}$ in (8) are negative and each row contains a nonzero entry, then the family $\mathcal{F}_{A, V}$ has multiple $S$-zeros and stop. If not, proceed.
6. Check whether the assumptions in Proposition 2.2 are fulfilled. If yes, the family $\mathcal{F}_{A, V}$ has multiple $S$-zeros and stop. If not, proceed.
7. Permute the columns of $A, V$ and the entries of $\kappa$ simultaneously such that the first $s \times s$ minor of $A$ is nonzero and go back to step 1.

We refer to the step 0. as the standard injectivity method, since it is the approach underlying the previous methods. For step 3. we use the determinant criterion if the rank of $A$ and the dimension of $S$ are appropriate. This is the case in the applications in the next section.

For step 7, we fix an order of the set of subsets of $[r]$ with $s$ elements. At the $i$-th iteration of the method, we consider the permutation that sends the $i$-th subset to the front, such that the set of the first $s$ indices and the set of last $r-s$ indices each remain ordered. We check whether the first $s \times s$ minor of $A$ is nonzero. Initial dependencies of the columns of $A$ can be taken into account beforehand, to reduce the number of checks. For example, if two columns of $A$ are linearly dependent, as is often the case for reaction networks, then subsets of $[r]$ with $s$ elements containing the two columns are disregarded.

Except for step 6, which is non-trivial, all the other steps are easily implemented using any mathematical software that allows for symbolic computations. We have automatised steps 1-5 and $\mathbf{7}$ in Maple and applied the method to a number of situations in the next section. Only when these steps were inconclusive, was it checked whether step $\mathbf{6}$ would give a positive answer (see Section (3).

## 3 Application to chemical reaction networks

The main application of the method is to determine whether a given chemical reaction network admits multiple steady states.

Setting. A (chemical) reaction network over a set $\mathcal{X}=\left\{X_{1}, \ldots, X_{n}\right\}$ is a finite collection of reactions

$$
\sum_{\ell=1}^{n} \mu_{\ell i} X_{\ell} \rightarrow \sum_{\ell=1}^{n} \beta_{\ell i} X_{\ell}, \quad i \in[r]
$$

where $\mu_{\ell i}, \beta_{\ell i} \in \mathbb{Z}_{\geq 0}$ and the two sides of a reaction are different. Let $A$ be the $n \times r$ matrix whose $(\ell, i)$-th entry is $\beta_{\ell i}-\mu_{\ell i}$, that is, the net production of $X_{\ell}$ in the $i$-th reaction.

The elements $X_{\ell}$ correspond to chemical species. We denote the concentration of $X_{\ell}$ by $x_{\ell}$. The vector of concentrations is $x=\left(x_{1}, \ldots, x_{n}\right) \in \mathbb{R}_{\geq 0}^{n}$ and the concentration at time $t$ is denoted by $x(t)$, although reference to time is often omitted from the notation.

It is custom to model the evolution of the concentrations in time using ODEs. A typical choice of ODE system is based on so-called mass-action kinetics, which is a special case of power-law kinetics. In this setting, the $j$-th reaction is assigned a vector $v^{j} \in \mathbb{R}^{n}$, and the monomial $x^{v^{j}}$ is assumed to be proportional to the rate of the $j$-th reaction when the system has concentration $x$.

Let $V$ be the $n \times r$ matrix with columns $v^{1}, \ldots, v^{r}$ and let $\kappa \in \mathbb{R}_{>0}^{r}$ be fixed constants. Then the system of ODEs is given by

$$
\begin{equation*}
\frac{d x}{d t}=f_{\kappa}(x), \quad f_{\kappa}(x)=A\left(\kappa \circ x^{V}\right) \tag{20}
\end{equation*}
$$

This ODE system is defined for $x \in \Omega_{V}$, where $\mathbb{R}_{>0}^{n} \subseteq \Omega_{V} \subseteq \mathbb{R}_{\geq 0}^{n}$ is obtained by removing from $\mathbb{R}_{\geq 0}^{n}$ the $\ell$-th hyperplane orthant whenever the $\ell$-th row of $\bar{V}$ contains a negative entry. The vector $\kappa$ is called the vector of reaction rate constants.

In mass-action kinetics, the matrix $V=\left(v_{i j}\right)$ is defined by $v_{\ell i}=\mu_{\ell i}$. That is, the $i$-th column of $V$ consists of the coefficients in the left-hand side of the $i$-th reaction.

The subspace $S=\operatorname{im}(A) \subseteq \mathbb{R}^{n}$ is called the stoichiometric subspace. Because the derivative of $x$ belongs to $S$, the trajectories of (20) are confined to sets $\left(x^{*}+S\right) \cap \mathbb{R}_{\geq 0}^{n}$ with $x^{*} \in$ $\mathbb{R}_{\geq 0}^{n}$ the initial condition of the system. These sets are called stoichiometric compatibility classes. We study therefore the dynamics of (20) confined to the stoichiometric compatibility classes and determine the steady states within each stoichiometric compatibility class. In this context, if the family $\mathcal{F}_{A, V}$ has multiple $S$-zeros. then one says that the reaction network is multistationary or admits multiple steady states.

Two-component systems. Consider the reaction network

$$
X_{1} \xrightarrow{\kappa_{1}} X_{2} \quad X_{2}+X_{3} \underset{\kappa_{3}}{\stackrel{\kappa_{2}}{\rightleftharpoons}} X_{1}+X_{4} \quad X_{4} \xrightarrow{\kappa_{4}} X_{3},
$$

where, as it is custom, reaction rate constants are written as reaction labels. This network models a simple bacterial two-component system in which $X_{1}, X_{2}$ (resp. $X_{3}, X_{4}$ ) are the unphosphorylated and phosphorylated forms of a histidine kinase (resp. response regulator). Using mass-action kinetics, the ODE system (20) of network (21) has matrices

$$
A=\left(\begin{array}{rrrr}
-1 & 1 & -1 & 0 \\
1 & -1 & 1 & 0 \\
0 & -1 & 1 & 1 \\
0 & 1 & -1 & -1
\end{array}\right), \quad V=\left(\begin{array}{llll}
1 & 0 & 1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 1
\end{array}\right)
$$

such that

$$
f_{\kappa}(x)=A\left(\kappa_{1} x_{1}, \kappa_{2} x_{2} x_{3}, \kappa_{3} x_{1} x_{4}, \kappa_{4} x_{4}\right)^{t}
$$

The family $\mathcal{F}_{A, V}$ is $S$-injective, and hence there is not a choice of constants for which a stoichiometric compatibility class has multiple steady states.

The matrix $A$ has rank 2, and notice that the second and fourth row are precisely the matrix $A$ in (5). Further, the vector subspace $S$ in (10) is precisely the stoichiometric subspace of this network, that is, the image of $A$. The matrix of exponents $V$ in (5) provides another kinetics for network (21), which is not mass-action. By the results above, network (21) with the kinetics given by $V$ is not multistationary.

On the other hand, we have shown that with $V$ given in (16), network (21) admits multiple steady states in one stoichiometric compatibility class for some choice of reaction rate constants $\kappa \in \mathbb{R}_{>0}^{4}$. Hence the network is multistationary.

Bifunctional kinase. Consider the following reaction network:

$$
\begin{array}{lll}
X_{1} \xrightarrow{\kappa_{1}} X_{2} & X_{2}+X_{3} \xlongequal[\kappa_{3}]{\kappa_{2}} X_{1}+X_{4} & X_{4}+X_{5} \xlongequal[\kappa_{5}]{\stackrel{\kappa_{4}}{\rightleftharpoons}} X_{3}+X_{6} \\
X_{4} \xrightarrow{\kappa_{8}} X_{3} & X_{6}+X_{7} \xlongequal[\kappa_{7}]{\kappa_{6}} X_{5}+X_{8} & X_{1}+X_{4} \xlongequal[\kappa_{11}]{\kappa_{10}} X_{9} \xrightarrow{\kappa_{12}} \\
X_{8} \xrightarrow{\kappa_{9}} X_{7} & &
\end{array}
$$

This network models a phosphorelay signalling system with bifunctional kinase 18. Using mass-action kinetics, the matrices $A, V$ in (20) are given as

$$
A=\left(\begin{array}{rrrrrrrrrrrr}
-1 & 1 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & -1 & 1 & 1 \\
1 & -1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & -1 & 1 & 1 & -1 & 0 & 0 & 1 & 0 & 0 & 0 & 1 \\
0 & 1 & -1 & -1 & 1 & 0 & 0 & -1 & 0 & -1 & 1 & 0 \\
0 & 0 & 0 & -1 & 1 & 1 & -1 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 1 & -1 & -1 & 1 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & -1 & 1 & 0 & 1 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 1 & -1 & 0 & -1 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & -1 & -1
\end{array}\right)
$$

and

$$
V=\left(\begin{array}{llllllllllll}
1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 1 & 1 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1
\end{array}\right)
$$

The family $\mathcal{F}_{A, V}$ is not $S$-injective. Since the rank of the stoichiometric matrix $A$ is 5 , we redefine $A$ to be the submatrix of $A$ consisting of rows $2,4,6,8,9$. Using steps 1-3, we conclude that the network does not admit multiple steady states. In other words, Gauss reduction of this submatrix of $A$ gives two matrices $A^{\prime}, V^{\prime}$ such that the family $\mathcal{F}_{A^{\prime}, V^{\prime}}$ is $S$-injective.

Apoptosis. We next consider a reaction network, which is a basic model of caspase activation for apoptosis [19]:

$$
\begin{array}{lcl}
X_{2}+X_{3} \xrightarrow{\kappa_{1}} X_{2}+X_{4} & 0 \xrightarrow{\kappa_{7}} X_{1} \xrightarrow{\kappa_{6}} 0 & X_{2} \xrightarrow{k_{12}} 0 \\
X_{1}+X_{4} \xrightarrow{\kappa_{2}} X_{2}+X_{4} & 0 \xrightarrow{\kappa_{9}} X_{3} \xrightarrow{\kappa_{8}} 0 & X_{4} \xrightarrow{k_{13}} 0 \\
X_{4}+X_{5} \xrightarrow{\kappa_{3}} X_{2}+X_{4} & 0 \xrightarrow{\kappa_{11}} X_{5} \xrightarrow{\kappa_{10}} 0 & X_{6} \xrightarrow{\kappa_{4}} 0 \\
X_{4}+X_{5} \xrightarrow{\kappa_{5}} X_{4} . & &
\end{array}
$$

With mass-action kinetics, this network is known to admit multiple steady states [19]. The rank of the stoichiometric subspace is maximal, that is, $S=\mathbb{R}^{6}$. For this network, steps 1-5 are inconclusive, for all possible column permutations, that is, we can neither assert nor reject that multiple steady states exist.

We permute the columns of $A, V$ such that the columns $7,9,11,12,13,14$ are first and apply Gauss reduction. The matrix $A_{1}$ in (8) becomes

$$
A_{1}=\left(\begin{array}{rrrrrrrr}
0 & -1 & 0 & 0 & 0 & -1 & 0 & 0 \\
-1 & 0 & 0 & 0 & 0 & 0 & -1 & 0 \\
0 & 0 & 0 & -1 & -1 & 0 & 0 & -1 \\
0 & -1 & 0 & 0 & 0 & 0 & 0 & 0 \\
-1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\
0 & 0 & -1 & 1 & 0 & 0 & 0 & 0
\end{array}\right)
$$

Observe that $A_{1}$ fulfils the assumptions in Proposition 2.1. We check step 6. The matrices $A^{\prime}, V^{\prime}$ are in this case

$$
A^{\prime}=\left(\begin{array}{rrrrrrrrrrrr}
1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & -1 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & -1
\end{array}\right)
$$

and

$$
V^{\prime}=\left(\begin{array}{rrrrrrrrrrrr}
1 & 1 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0 & 0 & 0 & -1 & 1 & 0 & 0 & 0 \\
0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\
1 & 0 & 0 & 0 & 0 & 1 & 0 & 1 & -1 & -1 & 1 & 0 \\
0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 & -1 & 0
\end{array}\right)
$$

The vector $\eta=(1,3,1,1,1,1,1,1,16 / 15,1,1,1 / 2)$ fulfils $A^{\prime} \eta \in \mathbb{R}_{>0}^{6}$ and that the determinant of $M_{\eta, \lambda}$ vanishes for all $\lambda \in \mathbb{R}_{>0}^{6}$. Therefore, by Proposition 2.2 and step 6 , we conclude that this network admits multiple steady states.

Networks of gene regulation. In [20] a total of 40,680 reaction networks modelling gene regulatory systems with mass-action kinetics are considered and analysed for multistationarity. The authors use the CRNT toolbox [8] together with a method termed network ancestry, justified by theoretical results in 21].

The authors determine that 2,654 out of the 40,680 networks cannot have multiple steady states. The standard injectivity method correctly precludes multistationarity for 691 of these 2,654 and is inconclusive for the remaining networks. Using the method presented here we can conclude that all 2,654 networks are not multistationary.

Interestingly, the authors cannot decide whether multiple steady states occur for 1,050 out of the 40,680 networks. The method successfully classifies these 1,050 networks and we can conclude that 47 of them are multistationary, while the remaining 1,003 cannot have multiple steady states. The remaining 36,976 networks are shown to be multistationary in [20]. The method is applied to the smallest 2,000 of these, and we reach the same conclusion.

Atoms of multistationarity. In [21], all possible networks consisting of two reactions and at most two molecules at each side of a reaction are considered. Here each reaction can be either irreversible $\longrightarrow$ or reversible $\rightleftharpoons$. So-called flow reactions $0 \rightleftharpoons X$ are added for each chemical species, such that $S=\mathbb{R}^{n}$. The authors determine which of these networks are multistationary when taken with mass-action kinetics.

We considered the 142 networks for which the standard injectivity criterion does not rule out multistationarity. Among these, the authors show that precisely 35 are multistationary and hence 107 are not multistationary. The new method does not perform as good as in the other test cases. Indeed, only 24 networks are identified as non-multistationary, while the remaining 118 are left unclassified.

A plausible explanation for the failure is the following. With mass-action kinetics, a reaction of the form $0 \rightarrow X$, called inflow, contributes a constant term $\kappa_{i}$ to the $i$-th component of $f_{\kappa}$ in (20). Therefore, we can write $f_{\kappa}(x)=\widetilde{f}_{\eta}(x)+\widetilde{\kappa}$, for some vector $\widetilde{\kappa} \in \mathbb{R}_{>0}^{n}$. It follows that $f_{\kappa}(x)=f_{\kappa}(y)$ if and only if $\widetilde{f}_{\eta}(x)=\widetilde{f}_{\eta}(y)$ and hence the standard injectivity method checks for injectivity of $\widetilde{f}_{\eta}$. In fact, $\widetilde{f}_{\eta}$ is $g_{\widehat{\kappa}}$ in our method, if the reactions are ordered such that the inflow reactions are first.

## 4 Discussion

Injectivity-based methods are used to preclude multistationarity in reaction networks [5, 6, 6, 9[16, 22]. The rationale behind the methods is that multiple zeros cannot occur if the map is injective.

In [6, 9, 14, 15, 22] the modelling framework is either mass-action or power-law kinetics, similar to what is used here. A common aspect of these works is that the mathematical development focuses on the vector subspace $S$ being the image of the coefficient matrix $A$.

In [5, 12], injectivity of a monomial map is studied in order to assert or preclude multistationarity. In both works, the authors are interested in determining whether a generalised binomial map, that is, a generalised polynomial map with two terms, admits multiple positive zeros. The positive zeros of a binomial map can be parameterised by a monomial map obtained by dividing one term of the binomial by the other term. In this way, there is passage from the non-existence of multiple positive zeros of the binomial map to the injectivity of the monomial map, which is identical to the passage from the study of $f_{\kappa}$ to the study of $g_{\hat{\kappa}}$ here.

The connection between [5, 12] and [6, 9, 14, 15] is clarified in [16], where unifying sign and determinant conditions for the injectivity of generalised polynomial maps are given. An important novelty of [16] is that $S$ is given independently of the image of $A$, in contrast to earlier work. This uncoupling is key in the results presented here.

We have demonstrated by examples that the new method can be applied to a vast amount of networks for which standard injectivity approaches are inconclusive. We have further applied the method to the five basic building blocks in cell signalling in [23] that are shown to admit multiple steady states. The method correctly classifies the networks as multistationary as well. However, we are not guaranteed that the method will classify any given network, as we discussed for the two-reaction networks in [21].

We focus on Gauss reduction of the generalised polynomial maps. Gauss reduction preserves the number of equations, which implies that the determinant criterion can be used to check step 3, when applying the method to reaction networks. Further, the method can be applied as a black box. However, the steps presented here also apply if we replace $f_{\kappa}$ by any set of generalised polynomial maps with the same positive zeros as $f_{\kappa}$. For example, if $f_{\kappa}$ is polynomial, one might consider a Gröbner basis of the system. In this sense, our work is a generalisation of [5] to the case where the Gröbner basis is not binomial. If the new set of equations differs from $f_{\kappa}$ in number of equations, then the sign criterion is to be used at step 3, which can also be computationally checked [16].

Finally, the method has been presented in connection with reaction networks. However,
the mathematical framework is given in full generality and can be used to obtain information on the number of positive zeros of generalised polynomial maps, independently of the context in which the question arises. In particular, since any polynomial can be embedded into a family $\mathcal{F}_{A, V}$, the method can be applied to preclude the existence of multiple positive solutions to any polynomial equation by letting $S=\mathbb{R}^{n}$.

## Acknowledgments

B. Joshi, A. Shiu, and D. Siegal-Gaskins are thanked for providing raw data and discussions of their results in 21 and [20. C. Wiuf is thanked for useful discussions and comments on earlier versions of this manuscript. My co-authors in [16] are thanked for interesting discussions on injectivity.

This work has been partially supported by project MTM2012-38122-C03-01/FEDER from the Ministerio de Economía y Competitividad, Spain, the Carlsberg Foundation, and the Lundbeck Foundation.

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[^0]:    E. Feliu (efeliu@math.ku.dk). Dept. of Mathematical Sciences, University of Copenhagen, Universitetsparken 5, 2100 Denmark

