Coupled cell networks: 
semigroups, Lie algebras and normal forms 

Bob Rink* and Jan Sanders†

March 3, 2014

Abstract
We introduce the concept of a semigroup coupled cell network and show that the collection of semigroup network vector fields forms a Lie algebra. This implies that near a dynamical equilibrium the local normal form of a semigroup network is a semigroup network itself. Networks without the semigroup property will support normal forms with a more general network architecture, but these normal forms nevertheless possess the same symmetries and synchronous solutions as the original network. We explain how to compute Lie brackets and normal forms of coupled cell networks and we characterize the SN-decomposition that determines the normal form symmetry. This paper concludes with a generalization to non-homogeneous networks with the structure of a semigroupoid.

1 Introduction
Coupled cell networks appear in many of the sciences and range from crystal models and electrical circuits to numerical discretization schemes, Josephson junction arrays, power grids, the world wide web, ecology, neural networks and systems biology. Not surprisingly, there exists an overwhelming amount of literature on coupled cell networks.

The last decade has seen the development of an extensive mathematical theory of dynamical systems with a network structure, cf. [RW], [SU], [TV], [UZ], [VR]. In these network dynamical systems, the evolution of the state of a constituent or “cell” is determined by the states of certain particular other cells. It is generally believed that a network structure has an impact on the behavior of a dynamical system, but it is not always clear how and why.

As an example, let us mention a system of differential equations with a homogeneous coupled cell network structure of the form

\[ \dot{x}_i = f(x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)}) \text{ for } 1 \leq i \leq N. \] (1.1)

These differential equations generate a dynamical system in which the evolution of the variable \( x_i \) is only determined by the values of \( x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)} \). The functions

\[ \sigma_1, \ldots, \sigma_n : \{1, \ldots, N\} \rightarrow \{1, \ldots, N\} \]

can thus be thought of as a network that prescribes how which cells influence which cells.

The literature on network dynamical systems focuses on the analysis of equilibria, periodic solutions, symmetry, synchrony, structural stability and bifurcations. As in the classical theory of dynamical systems, one often faces the task here of computing a local normal form near a dynamical equilibrium. These normal forms are obtained from coordinate transformations, and in their computation one calculates Lie brackets of vector fields, either implicitly or explicitly. It is here that one encounters an important technical problem:

Differential equations of the form (1.1) in general do not form a Lie algebra.

*Department of Mathematics, VU University Amsterdam, The Netherlands, b.w.rink@vu.nl.
†Department of Mathematics, VU University Amsterdam, The Netherlands, jan.sanders.a@gmail.com.
As a consequence one can not expect that the normal form of a coupled cell network is a coupled cell network as well. This complicates the local analysis and classification of network dynamical systems, because it means that one always has to compute the normal form of a network explicitly to understand its generic behavior - unless one is willing to assume that the network is given in normal form from the beginning, cf. [24], [29]. Normal form computations in [13], [21], [26] have revealed that a network structure can have a nontrivial impact on this generic behavior. One wants to understand and predict this.

In this paper, we will formulate an easily verifiable condition on a network structure under which the coupled cell network vector fields do form a Lie subalgebra of the Lie algebra of vector fields. Our main result is the following:

If \( \{ \sigma_1, \ldots, \sigma_n \} \) is a semigroup, then the differential equations (1.1) form a Lie algebra.

In this case, the local normal form of (1.1) is also of the form (1.1).

In addition, we show that the Lie bracket of semigroup coupled cell network vector fields can be lifted to a symbolic bracket that only involves the function \( f \). Normal form calculations can be performed at this symbolic level and one only returns to the reality of the differential equation when one is done computing. We also show that the symbolic space carries a dynamics of its own, determined by a certain fundamental network.

This situation is analogous to that of Hamiltonian vector fields, of which the Lie bracket is determined by the Poisson bracket of Hamiltonian functions. As a consequence, Hamiltonian normal forms are usually computed at the level of functions. Moreover, the symbolic dynamics of Hamiltonian functions is determined by a Poisson structure, cf. [41].

When \( \sigma_1, \ldots, \sigma_n \) do not form a semigroup, then we suggest that one simply completes them to the smallest collection

\[
\sigma_1, \ldots, \sigma_n, \sigma_{n+1}, \ldots, \sigma_{n'} : \{1, \ldots, N\} \to \{1, \ldots, N\}
\]

that does form a semigroup under composition. Then (1.1) can be written as

\[
\dot{x}_i = f'(x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)}) \quad \text{with} \quad f'(X_1, \ldots, X_n, X_{n+1}, \ldots, X_{n'}) := f(X_1, \ldots, X_n).
\]

The normal form of (1.1) will now lie within the extended class of semigroup coupled cell networks, there being no guarantee that it is again of the original form (1.1).

Thus one can choose: either to respect any given network structure as if it were a law of nature, so that no normal form can be computed, or to extend every network to a semigroup network and live with the consequences. One can object that even simple networks may need a lot of extension before they form a semigroup. But as an argument in favor of the semigroup approach, let us mention that the symmetries and synchrony spaces of a network are not at all affected by our semigroup extension. This implies in particular that these symmetries and synchrony spaces will also be present in the local normal form of the network. This latter property is both pleasant and important, if only in view of the large amount of research that has been devoted to symmetry [5], [6], [9], [18], [25], [31] and synchrony [2], [3], [4], [8], [10], [11], [27], [32], [34], [38], [49], [51], [53] in coupled cell networks. Semigroups may well be the natural invariants of coupled cell networks, even more than groups and symmetries.

Normal forms are computed by applying coordinate transformations [42], [45], [46], [47]. These transformations can be in the phase space of a differential equation, but in our case they take place in the space of functions \( f \) and have the form of a series expansion

\[
f \mapsto e^{a_0 g} f = f + \text{ad}_g^1(f) + \frac{1}{2} (\text{ad}_g^2)^2(f) + \ldots.
\]

Here \( f \) is the function to be transformed and normalized, \( g \) generates the coordinate transformation and \( \text{ad}_g \) denotes a representation, in this case the adjoint representation of the Lie algebra of \( f \)'s. Although at first sight this may seem a needlessly complicated way to describe coordinate transformations, this “Lie formalism” allows for a very flexible theory which streamlines both the theory and the computations.

The actual computation of the normal form of the function \( f \), and in particular the matter of solving homological equations, will not be entirely standard in the context of networks. Some things remain as in the theory of generic vector fields. For example, we show that the adjoint action of a linear element admits an SN-decomposition that determines a normal form symmetry. Other aspects may not carry through, such as the applicability of the Jacobson-Morozov lemma to characterize the complement of the image of the adjoint
action of a nilpotent element [7]. This is because the Lie algebra of the linear coupled cell network vector fields need not be reductive.

This paper is organized as follows. After giving a formal definition of a homogeneous coupled cell network in Section 2, we show in Section 3 that semigroups arise naturally in the context of coupled cell networks. In Sections 4 and 5 we prove that semigroup network dynamical systems are closed under taking compositions and Lie brackets. Section 6 explains how to compute the normal form of a network dynamical system, while in Sections 7 and 8 we prove that this normal form inherits both the symmetries and the synchrony spaces of the original network. In Section 9 we investigate the SN-decomposition of a linear coupled cell network vector field. This decomposition determines the normal form symmetry. Section 10 describes the aforementioned fundamental network. In Section 11 we actually compute the normal forms of some simple but interesting coupled cell networks, thus demonstrating that a coupled cell network structure can force anomalous steady state bifurcations. Finally, we show in Section 12 that our theory is also applicable to non-homogeneous or “colored” networks that display the structure of a semigroupoid.

Issues that we do not touch in this paper but aim to treat in subsequent work include:

1. The development of a linear algebra of semigroup coupled cell systems in order to define for example a “semigroup network Jordan normal form”.
2. Application of the results in this paper to semigroup networks that arise in applications, such as feed-forward motifs.
3. Better understanding the relation between semigroups and the well-known “groupoid formalism”. The results of this paper mainly apply to networks for which the groupoid is trivial and it would be interesting to study the impact of nontrivial “input symmetries” on bifurcations and normal forms.

1.1 Acknowledgement

We are grateful to Jan Bouwe van den Berg and Eddie Nijholt for carefully reading this manuscript at an early stage and for pointing out a number of mistakes and typos.

2 Homogeneous coupled cell networks

We shall be interested in dynamical systems with a coupled cell network structure. Such a structure can be determined in various ways [16], [35], [38], [51], but we choose to describe it here by means of a collection of distinct maps

$$\Sigma = \{\sigma_1, \ldots, \sigma_n\} \text{ with } \sigma_1, \ldots, \sigma_n : \{1, \ldots, N\} \to \{1, \ldots, N\}.$$ 

The collection $\Sigma$ has the interpretation of a network with $1 \leq N < \infty$ cells. Indeed, it defines a directed multigraph with $N$ vertices and precisely $n$ arrows pointing into each vertex, where the arrows pointing towards vertex $1 \leq i \leq N$ emanate from the vertices $\sigma_1(i), \ldots, \sigma_n(i)$. The number $n$ of incoming arrows per vertex is sometimes called the valence of the network.

In a network dynamical system we think of every vertex $1 \leq i \leq N$ in the network as a cell, of which the state is determined by a variable $x_i$ that takes values in a vector space $V$.

**Definition 2.1** Let $\Sigma = \{\sigma_1, \ldots, \sigma_n\}$ be a collection of $n$ distinct maps on $N$ elements, $V$ a finite dimensional real vector space and $f : V^n \to V$ a smooth function. Then we define

$$\gamma_f : V^N \to V^N \text{ by } (\gamma_f)_i(x) := f(x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)}) \text{ for } 1 \leq i \leq N. \quad (2.2)$$

Depending on the context, we will say that $\gamma_f$ is a homogeneous coupled cell network map or a homogeneous coupled cell network vector field subject to $\Sigma$. △

**Remark 2.2** In the literature, $\gamma_f$ is also called an admissible map/vector field. These maps and vector fields are commonly defined in terms of the groupoid of input equivalences of the cells in the network, see [35], [38] and [51]. For network maps/vector fields of the form (2.2) the corresponding groupoid is trivial though: every cell $i$ unambiguously has precisely one well-defined $j$-th input. A nontrivial groupoid arises when two or more different inputs of a cell can be interchanged without any effect on the dynamics. This case will be discussed in some detail in Section 8. △
Dynamical systems with a coupled cell network structure arise when we iterate the map $\gamma_f$ or integrate the vector field that it defines. The iterative dynamics on $V^N$ has the special property that the state of cell $i$ at time $m + 1$ depends only on the states of the cells $\sigma_1(i), \ldots, \sigma_n(i)$ at time $m$:

$$x^{(m+1)} = \gamma_f(x^{(m)}) \text{ if and only if } x_i^{(m+1)} = f(x_{\sigma_1(i)}^{(m)}, \ldots, x_{\sigma_n(i)}^{(m)}) \text{ for } 1 \leq i \leq N.$$  \hfill (2.3)

The continuous-time dynamical system on $V^N$ displays the same property infinitesimally: it is determined by the ordinary differential equations

$$\dot{x} = \gamma_f(x) \text{ if and only if } \dot{x}_i = f(x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)}) \text{ for } 1 \leq i \leq N.$$  \hfill (2.4)

We aim to understand how the network structure of $\gamma_f$ impacts these dynamical systems.

**Example 2.3** An example of a directed multigraph is shown in Figure 1, where the number of cells is $N = 3$ and valence is equal to $n = 2$. The maps $\sigma_1$ and $\sigma_2$ are given by

$$\sigma_1(1) = 1, \sigma_1(2) = 2, \sigma_1(3) = 3,$$
$$\sigma_2(1) = 1, \sigma_2(2) = 1, \sigma_2(3) = 2.$$  

![Figure 1: The collection $\{\sigma_1, \sigma_2\}$ depicted as a directed multigraph.](image)

A coupled cell network map/vector field subject to $\{\sigma_1, \sigma_2\}$ is of the form

$$\gamma_f(x_1, x_2, x_3) = (f(x_1, x_1), f(x_2, x_1), f(x_3, x_2)).$$  \hfill (2.5)

This network has obtained some attention [13], [24], [26], [29], [40] because it supports an anomalous codimension-one nilpotent double Hopf bifurcation when $\dim V = 2$.

**Example 2.4** In this example we let $\sigma_1, \sigma_2$ be as in Example 2.3 and we also define $\sigma_3$ as

$$\sigma_3(1) = 1, \sigma_3(2) = 1, \sigma_3(3) = 1.$$  

The network defined by $\{\sigma_1, \sigma_2, \sigma_3\}$ is depicted in Figure 2.

A coupled cell network map/vector field subject to $\{\sigma_1, \sigma_2, \sigma_3\}$ has the form

$$\gamma_g(x_1, x_2, x_3) = (g(x_1, x_1, x_1), g(x_2, x_1, x_1), g(x_3, x_2, x_1)).$$  \hfill (2.6)

We remark that this example is a generalization of Example 2.3: if $\gamma_f$ is as in Example 2.3 and if we define $g(X_1, X_2, X_3) := f(X_1, X_2)$, then $\gamma_g = \gamma_f$. In other words, (2.5) arises as a special case of (2.6).
3 Semigroups

A first and obvious difficulty that arises in the study of coupled cell network dynamical systems is that the composition \( \gamma_f \circ \gamma_g \) of two coupled cell network maps with an identical network structure may not have that same network structure.

Dynamically, this implies for example that the equation \( \gamma_f(x) = x \) for the steady states of \( \gamma_f \) and the equation \( (\gamma_f)^n(x) = x \) for its periodic solutions may have quite a different nature. We illustrate this phenomenon in the following example:

Example 3.1 Again, let \( N = 3 \) and let \( \sigma_1, \sigma_2, \sigma_3 \) be defined as in Examples 2.3 and 2.4. If

\[
\gamma_f(x_1, x_2, x_3) = (f(x_1, x_1), f(x_2, x_1), f(x_3, x_2)), \\
\gamma_g(x_1, x_2, x_3) = (g(x_1, x_1), g(x_2, x_1), g(x_3, x_2)),
\]

are coupled cell network maps subject to \( \{\sigma_1, \sigma_2\} \), then the composition

\[
(\gamma_f \circ \gamma_g)(x_1, x_2, x_3) = (f(g(x_1, x_1), g(x_1, x_1)), f(g(x_2, x_1), g(x_1, x_1)), f(g(x_3, x_2), g(x_2, x_1))
\]

in general is not a coupled cell network map subject to \( \{\sigma_1, \sigma_2\} \).

On the other hand, when \( \gamma_f \) and \( \gamma_g \) are network maps subject to \( \{\sigma_1, \sigma_2, \sigma_3\} \), i.e.

\[
\gamma_f(x_1, x_2, x_3) = (f(x_1, x_1, x_1), f(x_2, x_1, x_1), f(x_3, x_2, x_1)), \\
\gamma_g(x_1, x_2, x_3) = (g(x_1, x_1, x_1), g(x_2, x_1, x_1), g(x_3, x_2, x_1)),
\]

then it holds that

\[
(\gamma_f \circ \gamma_g)_1(x_1, x_2, x_3) = f(g(x_1, x_1, x_1), g(x_1, x_1, x_1), g(x_1, x_1, x_1)), \\
(\gamma_f \circ \gamma_g)_2(x_1, x_2, x_3) = f(g(x_2, x_1, x_1), g(x_1, x_1, x_1), g(x_1, x_1, x_1)), \\
(\gamma_f \circ \gamma_g)_3(x_1, x_2, x_3) = f(g(x_3, x_2, x_1), g(x_2, x_1, x_1), g(x_1, x_1, x_1)).
\]

This demonstrates that \( \gamma_f \circ \gamma_g \) is also a coupled cell network map subject to \( \{\sigma_1, \sigma_2, \sigma_3\} \).

Indeed, \( \gamma_f \circ \gamma_g = \gamma_h \), where

\[
h(X_1, X_2, X_3) = f(g(X_1, X_2, X_3), g(X_2, X_3, X_3), g(X_3, X_3, X_3)).
\]

\( \triangle \)

To understand when, in general, the composition of two coupled cell network maps is again a coupled cell network map, we compute that

\[
(\gamma_f \circ \gamma_g)_i(x) = f(\ldots, (\gamma_{g,j_1}(x), \ldots) = f(\ldots, g(x_{\sigma_2(j_1(i)), \ldots}, x_{\sigma_n(j_1(i)), \ldots}). \quad (3.7)
\]

The right hand side of (3.7) is an \( i \)-independent function of \( (x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)}) \) precisely when for all \( 1 \leq j_1, j_2 \leq n \) and all \( 1 \leq i \leq N \) it holds that \( \sigma_{j_1}(\sigma_{j_2}(i)) = \sigma_{j_2}(i) \) for some \( 1 \leq j_3 \leq n \). In other words, \( \gamma_f \circ \gamma_g \) is a coupled cell network map when \( \Sigma \) is a semigroup:

![Figure 2: The collection \( \{\sigma_1, \sigma_2, \sigma_3\} \) depicted as a directed multigraph.](image-url)
Definition 3.2 We say that \( \Sigma = \{ \sigma_1, \ldots, \sigma_n \} \) is a semigroup if for all \( 1 \leq j_1, j_2 \leq n \) there is a unique \( 1 \leq j_3 \leq n \) such that \( \sigma_{j_1} \circ \sigma_{j_2} = \sigma_{j_3} \). △

Viewing \( \Sigma \) as a directed multigraph, the condition that it is a semigroup just means that this directed multigraph is closed under the backward concatenation of arrows.

Of course, an arbitrary collection \( \Sigma = \{ \sigma_1, \ldots, \sigma_n \} \) need not be a semigroup. Even so, \( \Sigma \) generates a unique smallest semigroup

\[
\Sigma' = \{ \sigma_1, \ldots, \sigma_n, \sigma_{n+1}, \ldots, \sigma_{n'} \}
\]

that contains \( \Sigma \).

It is clear that every coupled cell network map \( \gamma_f \) subject to \( \Sigma \) is also a coupled cell network map subject to the semigroup \( \Sigma' \). Indeed, if we define

\[
f'(X_1, \ldots, X_n, X_{n+1}, \ldots, X_{n'}) := f(X_1, \ldots, X_n)
\]

then it obviously holds that

\[
(\gamma_f)_i(x) = f'(x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)}, x_{\sigma_{n+1}(i)}, \ldots, x_{\sigma_{n'}(i)}) = f(x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)}) = (\gamma_f)_i(x).
\]

We thus propose to augment \( \Sigma \) to the semigroup \( \Sigma' \) and to think of every coupled cell network map subject to \( \Sigma \) as a (special case of a) coupled cell network map subject to \( \Sigma' \).

Example 3.3 Again, let \( N = 3 \) and let \( \sigma_1, \sigma_2, \sigma_3 \) be defined as in Examples 2.3 and 2.4. It holds that \( \sigma_2^3 = \sigma_3 \), so the collection \( \{ \sigma_1, \sigma_2 \} \) is not a semigroup. On the other hand, one quickly computes that the composition table of \( \{ \sigma_1, \sigma_2, \sigma_3 \} \) is given by

<table>
<thead>
<tr>
<th>( \circ )</th>
<th>( \sigma_1 )</th>
<th>( \sigma_2 )</th>
<th>( \sigma_3 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \sigma_1 )</td>
<td>( \sigma_1 \sigma_2 \sigma_3 )</td>
<td>( \sigma_2 \sigma_3 )</td>
<td>( \sigma_3 \sigma_3 )</td>
</tr>
<tr>
<td>( \sigma_2 )</td>
<td>( \sigma_1 \sigma_2 \sigma_3 )</td>
<td>( \sigma_2 \sigma_3 )</td>
<td>( \sigma_3 \sigma_3 )</td>
</tr>
<tr>
<td>( \sigma_3 )</td>
<td>( \sigma_1 \sigma_2 \sigma_3 )</td>
<td>( \sigma_2 \sigma_3 )</td>
<td>( \sigma_3 \sigma_3 )</td>
</tr>
</tbody>
</table>

This shows that \( \{ \sigma_1, \sigma_2, \sigma_3 \} \) is closed under composition and hence is the smallest semigroup containing \( \{ \sigma_1, \sigma_2 \} \). △

4 Composition of network maps

To understand better how network maps behave under composition and in order to simplify our notation, let us define the maps

\[
\pi_i : V^N \to V^n \text{ by } \pi_i(x_1, \ldots, x_N) := (x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)}) \text{ for } 1 \leq i \leq N.
\]

This definition allows us to write (2.2) simply as

\[
(\gamma_f)_i := f \circ \pi_i.
\]

Expression (3.7) moreover turns into the formula

\[
(\gamma_f \circ \gamma_g)_i = f \circ (g \circ \pi_{\sigma_1(i)} \times \ldots \times g \circ \pi_{\sigma_n(i)}).
\]

The following technical result helps us write the right hand side of (4.9) in the form \( h \circ \pi_i \) for some function \( h : V^n \to V \), whenever \( \Sigma \) is a semigroup.

Theorem 4.1 Let \( \Sigma = \{ \sigma_1, \ldots, \sigma_n \} \) be a semigroup. Then for all \( 1 \leq j \leq n \) there exists a linear map \( A_{\sigma_j} : V^n \to V^n \) with the property that

\[
A_{\sigma_j} \circ \pi_i = \pi_{\sigma_j(i)} \text{ for all } 1 \leq i \leq N \text{ and all } 1 \leq j \leq n.
\]

Moreover, it holds that \( A_{\sigma_{j_1}} \circ A_{\sigma_{j_2}} = A_{\sigma_{j_1 \circ j_2}} \) for all \( 1 \leq j_1, j_2 \leq n \).

Proof: Because \( \Sigma \) is a semigroup, we can associate to each map \( \sigma_j \in \Sigma \) a unique map \( \tilde{\sigma}_j : \{1, \ldots, n\} \to \{1, \ldots, n\} \) defined via the formula \( \sigma_{\tilde{\sigma}_j(k)} = \sigma_j \circ \sigma_k \).

We now define the map \( A_{\sigma_j} : V^n \to V^n \) as

\[
A_{\sigma_j}(X_1, \ldots, X_n) := (X_{\sigma_{\tilde{\sigma}_j}(1)}, \ldots, X_{\sigma_{\tilde{\sigma}_j}(n)}).
\]
With this definition it holds that
\[(A_{\sigma_j} \circ \pi_i)(x) = A_{\sigma_j}(x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)}) = (x_{\sigma_{S_1}(i)}, \ldots, x_{\sigma_{S_n}(i)}) \]
\[= (x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)}) = \pi_{\sigma_j}(x) .\]

Remark ing moreover that
\[\sigma_{\bar{\sigma}_k(j_1)(j_2)} = \sigma_{\bar{\sigma}_k(j_1)} \circ \sigma_{j_2} = \sigma_k \circ \sigma_{j_1} \circ \sigma_{j_2} = \sigma_k \circ \sigma_{\bar{\sigma}_1(j_2)} = \sigma_{\bar{\sigma}_k(\bar{\sigma}_j(j_2))} ,\]
and hence that \(\bar{\sigma}_{\sigma_k(j_1)(j_2)} = \bar{\sigma}_k(\bar{\sigma}_j(j_2))\) for all \(1 \leq k \leq n\), we also find that
\[A_{\sigma_{j_1}} \circ A_{\sigma_{j_2}}(X_1, \ldots, X_n) = A_{\sigma_{j_1}}(X_{\bar{\sigma}_1(j_1)}, \ldots, X_{\bar{\sigma}_n(j_2)}) = (X_{\bar{\sigma}_{S_1}(j_1)(j_2)}, \ldots, X_{\bar{\sigma}_{S_n}(j_1)(j_2)}) \]
\[= (X_{\bar{\sigma}_1(\bar{\sigma}_j_1(j_2))}, \ldots, X_{\bar{\sigma}_n(\bar{\sigma}_j_1(j_2))}) = A_{\sigma_{j_1}(j_2)}(X_1, \ldots, X_n) = A_{\sigma_{j_1} \circ \sigma_{j_2}}(X_1, \ldots, X_n) .\]

This proves the theorem. □

The identity \(A_{\sigma_{j_1}} \circ A_{\sigma_{j_2}} = A_{\sigma_{j_1} \circ \sigma_{j_2}}\) expresses that the \(A_{\sigma_j}\) form a representation of the semigroup \(\Sigma\). Using this representation we obtain:

**Theorem 4.2** Let \(\Sigma = \{\sigma_1, \ldots, \sigma_n\}\) be a semigroup. Define for \(f, g : V^n \to V\) the function
\[f \circ \Sigma g : V^n \to V\] by \(f \circ \Sigma g := f \circ ((g \circ A_{\sigma_1}) \times \ldots \times (g \circ A_{\sigma_n})) .\)

Then
\[\gamma f \circ \gamma g = \gamma (f \circ \Sigma g) .\]

**Proof:** From formula (4.9) and Theorem 4.1. □

Theorem 4.2 reveals once more that if \(\Sigma\) is a semigroup, then the composition of two coupled cell network maps \(\gamma_f\) and \(\gamma_g\) is again a coupled cell network map, namely \(\gamma (f \circ \Sigma g)\). More importantly, it shows how to compute \(f \circ \Sigma g\) "symbolically", i.e. using only the functions \(f\) and \(g\) and a representation of the network semigroup.

The final result of this section ensures that the "symbolic composition" \(\circ \Sigma\) makes the space \(C^\infty(V^n, V)\) into an associative algebra.

**Lemma 4.3**
\[(f \circ \Sigma g) \circ \Sigma h = f \circ \Sigma (g \circ \Sigma h) .\]

**Proof:**
\[(f \circ \Sigma g) \circ \Sigma h(X) = (f \circ \Sigma g)\ldots, h(A_{\sigma_k}X),\ldots) = f\ldots, g(A_{\sigma_j}(\ldots, h(A_{\sigma_k}X),\ldots),\ldots) = f\ldots, g\ldots, h(A_{\sigma_k}(\sigma_jX),\ldots) = f\ldots, g\ldots, h(A_{\sigma_k}A_{\sigma_j}X),\ldots) = f\ldots, (g \circ \Sigma h)(A_{\sigma_j}X),\ldots) = f \circ \Sigma (g \circ \Sigma h)(X) .\]

□

With Lemma 4.3 at hand, Theorem 4.2 just means that the linear map
\[\gamma : C^\infty(V^n, V) \to C^\infty(V^N, V_N)\] that sends \(f\) to \(\gamma_f\)
is a homomorphism of associative algebras.

**Example 4.4** Again, let \(N = 3\) and let \(\sigma_1, \sigma_2, \sigma_3\) be defined as in Examples 2.3 and 2.4. We recall that the composition table of \(\{\sigma_1, \sigma_2, \sigma_3\}\) was given in Example 3.3. The rows of this table express that
\[\begin{align*}
\bar{\sigma}_1(1) &= 1, \bar{\sigma}_1(2) = 2, \bar{\sigma}_1(3) = 3, \\
\bar{\sigma}_2(1) &= 2, \bar{\sigma}_2(2) = 3, \bar{\sigma}_2(3) = 3, \\
\bar{\sigma}_3(1) &= 3, \bar{\sigma}_3(2) = 3, \bar{\sigma}_3(3) = 3 .
\end{align*}\]
This implies in particular that
\[ A_{\sigma_{1}}(X_{1}, X_{2}, X_{3}) = (X_{1}, X_{2}, X_{3}), \]
\[ A_{\sigma_{2}}(X_{1}, X_{2}, X_{3}) = (X_{2}, X_{3}, X_{3}), \]
\[ A_{\sigma_{3}}(X_{1}, X_{2}, X_{3}) = (X_{3}, X_{3}, X_{3}). \]

Substitution in (4.13) therefore yields that
\[ f \circ \Sigma g (X_{1}, X_{2}, X_{3}) = f(g(X_{1}, X_{2}, X_{3}), g(X_{2}, X_{3}, X_{3}), g(X_{3}, X_{3}, X_{3})). \]

We conclude that \( f \circ \Sigma g \) equals the function \( h \) found in Example 3.1. \( \triangle \)

**Remark 4.5** The defining relation
\[ \sigma_{\tilde{\sigma}_{j}}(k) = \sigma_{j} \circ \sigma_{k} \text{ for } \sigma_{j} : \{1, \ldots, n\} \rightarrow \{1, \ldots, n\} \]
expresses that the map \( \tilde{\sigma}_{j} \) describes the left-multiplicative behavior of \( \sigma_{j} \). The computation
\[ \sigma_{\tilde{\sigma}_{j} \circ \tilde{\sigma}_{j}}(k) = \sigma_{j} \circ \sigma_{j} \circ \sigma_{k} = \sigma_{j} \circ \sigma_{\tilde{\sigma}_{j}}(k) = \sigma_{\tilde{\sigma}_{j}}(\tilde{\sigma}_{j}(k)) \]
moreover reveals that
\[ \tilde{\sigma}_{j} \circ \tilde{\sigma}_{j} = \tilde{\sigma}_{j} \circ \tilde{\sigma}_{j} \text{ for all } 1 \leq j_1, j_2 \leq n. \]
This means in particular that the collection \( \{\tilde{\sigma}_{1}, \ldots, \tilde{\sigma}_{n}\} \) is closed under composition.

The maps \( \tilde{\sigma}_{1}, \ldots, \tilde{\sigma}_{n} \) will play an interesting role in this paper. In fact, we will show in Section 10 that they are themselves the network maps of a certain “fundamental network” that fully determines the fate of all network dynamical systems subject to \( \Sigma. \) \( \triangle \)

**Remark 4.6** For a map \( \sigma : \{1, \ldots, n\} \rightarrow \{1, \ldots, m\} \) let us denote by \( \lambda_{\sigma} : V^{m} \rightarrow V^{n} \) the linear map
\[ \lambda_{\sigma}(X_{1}, X_{m}) := (X_{\sigma(1)}, \ldots, X_{\sigma(n)}). \]
This means that the matrix of the map \( \lambda_{\sigma} \) has precisely one \( \text{id}_{V} \) on each row and zeroes elsewhere. We will denote the space of such maps by
\[ \Lambda(m, n) := \{\lambda_{\sigma} : V^{m} \rightarrow V^{n} \mid \sigma : \{1, \ldots, n\} \rightarrow \{1, \ldots, m\}\}. \]
One quickly checks that the assignment \( \lambda : \sigma \mapsto \lambda_{\sigma} \) is contravariant. More precisely, if \( \sigma : \{1, \ldots, n\} \rightarrow \{1, \ldots, m\} \) and \( \tau : \{1, \ldots, m\} \rightarrow \{1, \ldots, l\} \), then
\[ \lambda_{\sigma} \circ \lambda_{\tau} = \lambda_{\tau \circ \sigma}. \]
In particular, \( \Lambda(m, m) \) is a semigroup and the map \( \sigma \mapsto \lambda_{\sigma} \) an anti-homomorphism from the semigroup of all maps from \( \{1, \ldots, m\} \) to itself to the semigroup \( \Lambda(m, m) \).

The maps \( \pi_{i} \) and \( A_{\sigma_{j}} \) defined above are examples of such maps:
\[ \pi_{i} = \lambda_{\sigma_{i}}, \text{ with } \sigma_{i} : \{1, \ldots, n\} \rightarrow \{1, \ldots, N\} \text{ defined as } \sigma_{i}(j) := \sigma_{j}(i). \]
\[ A_{\sigma_{j}} = \lambda_{\tilde{\sigma}_{j}}, \text{ with } \tilde{\sigma}_{j} : \{1, \ldots, n\} \rightarrow \{1, \ldots, n\} \text{ defined by } \tilde{\sigma}_{j}(k) := \tilde{\sigma}_{j}(k). \]
This observation can be used give a proof of Theorem 4.1 that is free of coordinates:

**Proof** [Of Theorem 4.1 without coordinates]: We observe that
\[ \sigma_{i}(\tilde{\sigma}_{j}(k)) = \sigma_{i}(\tilde{\sigma}_{k}(j)) = \sigma_{\tilde{\sigma}_{k}}(j)(i) = (\sigma_{k} \circ \sigma_{j})(i) = \sigma_{k}(\sigma_{j}(i)) = \sigma^{\sigma_{j}(i)}(k). \]
and hence that \( \sigma^{i} \circ \tilde{\sigma} = \sigma^{\sigma_{j}(i)}(i). \) Using this, we find that
\[ A_{\sigma_{j}} \circ \pi_{i} = \lambda_{\tilde{\sigma}_{j}} \circ \lambda_{\sigma_{i}} = \lambda_{\sigma_{j} \circ \tilde{\sigma}_{i}} = \lambda_{\sigma_{j}(i)} = \pi_{\sigma_{j}(i)}. \]
Similarly, the computation
\[ \sigma_{\tilde{\sigma}_{j} \circ \tilde{\sigma}_{j}}(k) = \sigma_{\tilde{\sigma}_{j}}(\tilde{\sigma}_{j}(k)) = \sigma_{\tilde{\sigma}_{j}}(\sigma_{j}(i)) = \sigma_{\tilde{\sigma}_{j}}(\sigma_{j}(i)) = \sigma_{\tilde{\sigma}_{j} \circ \tilde{\sigma}_{j}}(k) \]
\[ = \sigma_{\tilde{\sigma}_{j}(j)} \circ \sigma_{j} = \tilde{\sigma}_{j} \circ \sigma_{j} = \sigma_{j} \circ \tilde{\sigma}_{j}(k) = \sigma_{j}(\tilde{\sigma}_{j}(k)) = \sigma_{\tilde{\sigma}_{j} \circ \tilde{\sigma}_{j}}(k). \]
reveals that
\[ \widetilde{\sigma}^{i_2} \circ \widetilde{\sigma}^{j_1} = \widetilde{\sigma}^{\delta_{j_1}(j_2)} . \]
As a consequence,
\[ A_{\delta_j} \circ A_{\delta_2} = \lambda_{\delta_j} \circ \lambda_{\delta_2} = \lambda_{\delta_2 \circ \delta_j} = \lambda_{\delta_{\delta_j}(j_2)} = A_{\delta_j} \circ A_{\delta_2} . \]
\[ \square \]

Unfortunately, this coordinate free proof of Theorem 4.1 is relatively long. \[ \triangle \]

5 A coupled cell network bracket

We will now think of \( \gamma_f : V^N \to V^N \) as a vector field that generates the differential equation
\[ \dot{x} = \gamma_f(x) . \]
We suggestively denote by \( e^{t \gamma_f} \) the time-\( t \) flow of the vector field \( \gamma_f \) on \( V^N \) and by \( (e^{t \gamma_f})_* \gamma_f \) the pushforward of the vector field \( \gamma_f \) under the time-\( t \) flow of \( \gamma_f \). We recall that the Lie bracket of \( \gamma_f \) and \( \gamma_g \) is then the vector field \( [\gamma_f, \gamma_g] : V^N \to V^N \) defined as
\[ [\gamma_f, \gamma_g](x) := \left. \frac{d}{dt} \right|_{t=0} (e^{t \gamma_f})_* \gamma_g = D\gamma_f(x) \cdot \gamma_g(x) - D\gamma_g(x) \cdot \gamma_f(x) . \] (5.14)
The main result of this section is that if \( \Sigma \) is a semigroup, then the collection of coupled cell network vector fields is closed under taking Lie brackets.

**Theorem 5.1** Let \( \Sigma = \{\sigma_1, \ldots, \sigma_n\} \) be a semigroup and let the \( A_{\sigma_j} : V^n \to V^n \) be as in Theorem 4.1. Define, for \( f, g : V^n \to V \), the function \( [f, g]_{\Sigma} : V^n \to V \) by
\[ [f, g]_{\Sigma} := \sum_{j=1}^n D_j f \cdot (g \circ A_{\sigma_j}) - D_j g \cdot (f \circ A_{\sigma_j}) . \] (5.15)
Then
\[ [\gamma_f, \gamma_g] = [\gamma_f, \gamma_g]_{\Sigma} . \] (5.16)

**Proof:** We start by remarking that
\[ \gamma_f(x + t \gamma_g(x)) = f(\ldots, x_{\sigma_j(i)} + t \gamma_g(x)_{\sigma_j(i)}; \ldots) = f(\ldots, x_{\sigma_j(i)} + t g(x_{\sigma_1(\sigma_j(i))}, \ldots, x_{\sigma_n(\sigma_j(i))}); \ldots) = f(\ldots, x_{\sigma_j(i)} + t g(A_{\sigma_j}(x_{\sigma_1(\sigma_j(i))}, \ldots, x_{\sigma_n(\sigma_j(i))})); \ldots) . \]
Differentiating this identity with respect to \( t \) and evaluating the result at \( t = 0 \) gives that
\[ (D\gamma_f(x) \cdot \gamma_g(x))_i = \sum_{j=1}^n D_j f(x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)}) g(A_{\sigma_j}(x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)})) . \]
This proves that \( (D\gamma_f \cdot \gamma_g)_i = \left( \sum_{j=1}^n D_j f \cdot (g \circ A_{\sigma_j}) \right) \circ \pi_i \), and hence that \( D\gamma_f \cdot \gamma_g \) is a coupled cell network vector field. With a similar computation for \( D\gamma_g \cdot \gamma_f \), we thus find that the Lie bracket between \( \gamma_f \) and \( \gamma_g \) is given by
\[ [\gamma_f, \gamma_g]_i = (D\gamma_f \cdot \gamma_g)_i - (D\gamma_g \cdot \gamma_f)_i = \sum_{j=1}^n [D_j f \cdot (g \circ A_{\sigma_j}) - D_j g \cdot (f \circ A_{\sigma_j})] \circ \pi_i = [f, g]_{\Sigma} \circ \pi_i . \]
This proves the theorem. \[ \square \]

Lemma 5.2 below states that the “symbolic bracket” \([\cdot, \cdot]_{\Sigma}\) is a Lie bracket.

**Lemma 5.2** The bracket \([\cdot, \cdot]_{\Sigma}\) makes \( C^\infty(V^n, V) \) a Lie algebra. Moreover, the linear map \( \gamma : C^\infty(V^n, V) \to C^\infty(V^N, V^N) \) that sends \( f \) to \( \gamma_f \) is a Lie algebra homomorphism.
Proof: Anti-symmetry of $[\cdot , \cdot ]_\Sigma$ is clear from formula (5.15). The Jacobi identity

$$[f, [g, h]]_\Sigma + [g, [h, f]]_\Sigma + [h, [f, g]]_\Sigma = 0$$

follows from a somewhat lengthy computation as follows. First of all, because

$$A_\sigma_1, \ldots, h(A_{\sigma_k} X), \ldots) = (\ldots, h(A_{\sigma_k} A_\sigma(j) X), \ldots) = (\ldots, h(A_{\sigma_k} A_\sigma(j) X), \ldots),$$

we find that

$$(g \circ A_\sigma) (X + t(\ldots , h(A_{\sigma_k} X), \ldots)) = g(A_\sigma X + tA_\sigma(\ldots , h(A_{\sigma_k} X), \ldots))$$
$$= g(A_\sigma X + t(\ldots, h(A_{\sigma_k} A_\sigma(j) X), \ldots)).$$

Differentiating this identity with respect to $t$ and evaluating the result at $t = 0$ yields that

$$\sum_{k=1}^n D_k (g \circ A_\sigma) \cdot (h \circ A_{\sigma_k}) = \sum_{k=1}^n (D_k g \circ A_\sigma) \cdot (h \circ A_{\sigma_k} \circ A_\sigma).$$

With this in mind, we now compute that

$$[f, [g, h]]_\Sigma = \sum_{j,k=1}^n D_j f \cdot ([g, h]_\Sigma \circ A_\sigma) - D_j [g, h]_\Sigma \cdot (f \circ A_\sigma)$$
$$= \sum_{j,k=1}^n D_j f \cdot (D_k g \circ A_\sigma) \cdot (h \circ A_{\sigma_k} \circ A_\sigma) - D_j f \cdot (D_k h \circ A_\sigma) \cdot (f \circ A_{\sigma_k} \circ A_\sigma)
- D_k g \cdot D_j (h \circ A_{\sigma_k} \circ A_\sigma) \cdot (f \circ A_\sigma)
- D_k h \cdot D_j (g \circ A_{\sigma_k} \circ A_\sigma) \cdot (f \circ A_\sigma)
- D_k^2 g \cdot (h \circ A_{\sigma_k} \circ A_\sigma) \cdot (f \circ A_\sigma)_j
- D_k^2 h \cdot (g \circ A_{\sigma_k} \circ A_\sigma) \cdot (f \circ A_\sigma)_j.$$

Using the symmetry of the second derivatives, the Jacobi identity follows from cyclically permuting $f, g$ and $h$ in the above expression and summing the results. This proves that $C^\infty(V^n, V)$ is a Lie algebra. Theorem 5.1 means that $\gamma$ is a Lie algebra homomorphism. □

Example 5.3 Again, let $N = 3$ and let $\sigma_1, \sigma_2, \sigma_3$ be defined as in Examples 2.3 and 2.4. We recall that $A_{\sigma_1}, A_{\sigma_2}$ and $A_{\sigma_3}$ were computed in Example 4.4. It follows that

$$[f, g]_\Sigma(X) = D_1 f(X_1, X_2, X_3) \cdot g(X_1, X_2, X_3) + D_2 f(X_1, X_2, X_3) \cdot g(X_2, X_3, X_1)
+ D_3 f(X_1, X_2, X_3) \cdot g(X_3, X_1, X_2) - D_1 g(X_1, X_2, X_3) \cdot f(X_1, X_2, X_3)
- D_2 g(X_1, X_2, X_3) \cdot f(X_2, X_3, X_1) - D_3 g(X_1, X_2, X_3) \cdot f(X_3, X_1, X_2).$$

△

6 Coupled cell network normal forms

Normal forms are an essential tool in the study of the dynamics and bifurcations of maps and vector fields near equilibria, cf. [42], [47]. In this section we will show that it can be arranged that the normal form of a semigroup coupled cell network is a coupled cell network as well. This normal form can moreover be computed "symbolically", i.e. at the level of the function $f$. With Theorem 5.1 at hand, this result is perhaps to be expected. We nevertheless state two illustrative theorems in this section, and sketch their proofs.

We start by making a few standard definitions. First of all, we define for $f \in C^\infty(V^n, V)$ the operator $ad_{f}^{\Sigma} : C^\infty(V^n, V) \to C^\infty(V^n, V)$ by

$$ad_{f}^{\Sigma}(g) := [f, g]_\Sigma.$$ 

Next, we define for every $k = 0, 1, 2, \ldots$ the finite dimensional subspace

$$P^k := \{ f : V^n \to V \text{ homogeneous polynomial of degree } k + 1 \} \subset C^\infty(V^n, V).$$ 

(6.17)
One can observe that $P^0 = L(V^N, V)$ and that if $f \in P^k$ and $g \in P^l$, then $[f, g]_\Sigma \in P^{k+l}$, as is obvious from formula (5.15). In particular, we have that
\[
    \text{if } f_0 \in L(V^N, V) \text{ then } \text{ad}^\Sigma_{f_0} : P^k \to P^k.
\]

With this in mind, we formulate the first main result of this section. It essentially states that one may restrict the study of semigroup coupled cell networks near local equilibria to semigroup coupled cell networks of a very specific "normal form".

**Theorem 6.1 (Coupled cell network normal form theorem)** Let $\Sigma = \{\sigma_1, \ldots, \sigma_n\}$ be a semigroup, $f \in C^\infty(V^N, V)$ and assume that $f(0) = 0$. We Taylor expand $f$ as
\[
    f = f_0 + f_1 + f_2 + \ldots \text{ with } f_k \in P^k.
\]

Let $1 \leq r < \infty$ and for every $1 \leq k \leq r$, let $N^k \subset P^k$ be a subspace such that
\[
    N^k \oplus \text{im } \text{ad}^\Sigma_{f_0}\big|_{P^k} = P^k.
\]

Then there exists an analytic diffeomorphism $\Phi$, sending an open neighborhood of $0$ in $V^N$ to an open neighborhood of $0$ in $V^N$, that conjugates the coupled cell network vector field $\gamma_f$ to a coupled cell network vector field $\gamma_\overline{f}$ with
\[
    \overline{f} = f_0 + \overline{f}_1 + \overline{f}_2 + \ldots \text{ and } \overline{f}_k \in N^k \text{ for all } 1 \leq k \leq r.
\]

**Proof:** [Sketch] We only sketch a proof without estimates here, because the construction of a normal form by means of "Lie transformations" is very well-known.

For $g \in C^\infty(V^N, V)$ with $g(0) = 0$, the time-$t$ flow $e^{t\gamma_g}$ defines a diffeomorphism of some open neighborhood of $0$ in $V^N$ to another open neighborhood of $0$ in $V^N$. Thus we can consider, for any $f \in C^\infty(V^N, V)$, the curve $t \mapsto (e^{t\gamma_f})_* \gamma_f \in C^\infty(V^N, V^N)$ of pushforward vector fields. This curve satisfies the linear differential equation
\[
    \frac{d}{dt}(e^{t\gamma_g})_* \gamma_f = \frac{d}{dt}(e^{h\gamma_f})_* (e^{t\gamma_g})_* \gamma_f = [\gamma_g, (e^{t\gamma_g})_* \gamma_f] = \text{ad}_{\gamma_g}((e^{t\gamma_g})_* \gamma_f), \tag{6.18}
\]
where the second equality holds by definition of the Lie bracket of vector fields (5.14) and we have used the conventional definition of $\text{ad}_{\gamma_g} : C^\infty(V^N, V^N) \to C^\infty(V^N, V^N)$ as
\[
    \text{ad}_{\gamma_g}(\gamma_f) := [\gamma_g, \gamma_f].
\]

Solving the linear differential equation (6.18) together with the initial condition $(e^{0\gamma_g})_* \gamma_f = \gamma_f$, we find that the time-$1$ flow of $\gamma_g$ transforms $\gamma_f$ into
\[
    (e^{\gamma_g})_* \gamma_f = e^{\text{ad}_{\gamma_g}(\gamma_f)} = \gamma_f + [\gamma_g, \gamma_f] + \frac{1}{2}[[\gamma_g, \gamma_g], \gamma_f] + \ldots.
\]

The main point of this proof is that by Theorem 5.1 the latter expression is also equal to
\[
    \gamma_{f + [g, f]_\Sigma + \frac{1}{2}[g, [g, f]]_\Sigma + \ldots} = e^{\text{ad}_{\gamma_g}(\gamma_f)}.
\]

The diffeomorphism $\Phi$ in the statement of the theorem is now constructed as the composition of a sequence of time-$1$ flows $e^{\gamma_g}$ ($1 \leq k \leq r$) of coupled cell network vector fields $\gamma_{g_k}$ with $g_k \in P^k$. We first take $g_1 \in P^1$, so that $\gamma_f$ is transformed by $e^{\gamma_{g_1}}$ into
\[
    (e^{\gamma_{g_1}})_* \gamma_f = \gamma_{e^{\text{ad}_{\gamma_{g_1}}}}(\gamma_f) = \gamma_{f_0 + f_1 + f_2 + \ldots}
\]
in which
\[
    f_1^1 = f_1 + [g_1, f_0]_\Sigma \in P^1 \quad f_2^1 = f_2 + [g_1, f_1]_\Sigma + \frac{1}{2}[g_1, [g_1, f_0]]_\Sigma \in P^2 \\
    f_3^1 = f_3 + \ldots \in P^3 \text{ etc.}
\]

It is the fact that $N^1 \oplus \text{im } \text{ad}^\Sigma_{f_0}\big|_{P^1} = P^1$ that allows us to choose a (not necessarily unique) $g_1 \in P^1$ in such a way that
\[
    f_1^1 = f_1 + [g_1, f_0]_\Sigma = f_1 - \text{ad}^\Sigma_{f_0}(g_1) \in N^1.
\]
We proceed by choosing $g_2 \in P^2$ in such a way that $(e^{g_2} \circ e^{g_1}) \ast \gamma_f = (e^{g_2} \circ ((e^{g_1}) \ast \gamma_f)) = \gamma_{f_0 + f_1^2 + f_2^3 + \ldots}$ with $f_2^3 \in \mathbb{N}^2$. Continuing in this way, after $r$ steps we obtain that

$$\Phi := e^{g_r} \circ \ldots \circ e^{g_1}$$

transforms $\gamma_f$ into $\Phi \ast \gamma_f = \gamma_f \ast \gamma_0 + \gamma_1 + \ldots$ where $f_k^k \in \mathbb{N}^k$ for all $1 \leq k \leq r$.

Being the composition of finitely many flows of polynomial coupled cell network vector fields, $\Phi$ is obviously analytic.

In applications, one is often interested in the bifurcations that occur in the dynamics of a map or differential equation under the variation of external parameters. In the case of coupled cell networks, we may for example assume that $f \in C^\infty(V^n \times \mathbb{R}^p, V)$ and let

$$f^\lambda(X) := f(X; \lambda)$$

define a smooth parameter family in $C^\infty(V^n, V)$. Correspondingly, the coupled cell networks $\gamma_{f^\lambda}$ form a smooth parameter family in $C^\infty(V^N, V^N)$.

To formulate an appropriate normal form theorem for parameter families of coupled cell networks, we define for $k \geq -1$ and $l \geq 0$,

$$P^{k,l} := \{ f : V^n \times \mathbb{R}^p \to V \text{ homogeneous polynomial of degree } k+1 \text{ in } X \text{ and degree } l \in \lambda \}.$$ 

We observe that

$$[P^{k,l}, P^{K,L}]_\Sigma \subset P^{k+K,l+L},$$

which leads to the following

**Theorem 6.2 (Coupled cell network normal form theorem with parameters)** Let $\Sigma = \{\sigma_1, \ldots, \sigma_n\}$ be a semigroup, $f \in C^\infty(V^n \times \mathbb{R}^p, V)$ and $f(0; 0) = 0$. We Taylor expand

$$f = (f_{-1,1} + f_{-1,2} + \ldots) + (f_{0,0} + f_{0,1} + f_{0,2} + \ldots) + (f_{1,0} + f_{1,1} + f_{1,2} + \ldots) + \ldots$$

with $f_{k,l} \in P^{k,l}$.

Let $1 \leq r_1, r_2 < \infty$ and for every $-1 \leq k \leq r_1$ and $0 \leq l \leq r_2$, let $N^{k,l} \subset P^{k,l}$ be a subspace such that

$$N^{k,l} \oplus \text{im } \text{ad}^{\Sigma}_{f_{0,0}} |_{P^{k,l}} = P^{k,l}.$$ 

Then there exists a polynomial family $\Phi^\lambda$ of analytic diffeomorphisms, defined for $\lambda$ in an open neighborhood of 0 in $\mathbb{R}^p$ and each sending an open neighborhood of 0 in $V^N$ to an open neighborhood of 0 in $V^N$, with the property that $\Phi^\lambda$ conjugates $\gamma_{f^\lambda}$ to $\gamma_{\lambda^\ast}$, where

$$\bar{f} = (\bar{f}_{-1,1} + \bar{f}_{-1,2} + \ldots) + (\bar{f}_{0,0} + \bar{f}_{0,1} + \bar{f}_{0,2} + \ldots) + (\bar{f}_{1,0} + \bar{f}_{1,1} + \bar{f}_{1,2} + \ldots) + \ldots$$

and

$$\bar{f}_{k,l} \in N^{k,l} \text{ for all } -1 \leq k \leq r_1 \text{ and } 0 \leq l \leq r_2.$$ 

**Proof:** [Sketch] The procedure of normalization is similar as in the proof of Theorem 6.1. With respect to $\text{ad}^{\Sigma}_{f_{0,0}}$, one consecutively normalizes

$$f_{1,0}, f_{2,0}, \ldots, f_{r_2,0}; f_{-1,1}, f_{0,1}, f_{1,1}, \ldots, f_{r_1,1};$$

$$f_{-1,2}, f_{0,2}, f_{1,2}, \ldots, f_{r_2,2}; f_{-1,3}, f_{0,3}, f_{1,3}, \ldots, f_{r_2,3}.$$ 

Because $[P^{k,l}, P^{K,L}]_\Sigma \subset P^{k+K,l+L}$, we see that once $f_{k,l}$ has been normalized to $\bar{f}_{k,l}$, it is not changed/affected anymore by any of the subsequent normalization transformations.

Of course, Theorem 5.1 implies that many other standard results from the theory of normal forms will have a counterpart in the context of semigroup coupled cell networks as well.

We will compute the normal forms of some network differential equations in Section 11.

**Example 6.3** Again, let $N = 3$ and let $\sigma_1, \sigma_2, \sigma_3$ be defined as in Examples 2.3 and 2.4. If

$$\gamma_f(x_1, x_2, x_3) = (f(x_1, x_1), f(x_2, x_1), f(x_3, x_2))$$

is a coupled cell network subject to $\{\sigma_1, \sigma_2\}$, then its normal form will in general be a network subject to $\{\sigma_1, \sigma_2, \sigma_3\}$, i.e.

$$\gamma_{\bar{f}}(x_1, x_2, x_3) = \gamma_f(x_1, x_2, x_3) = (\bar{f}(x_1, x_1), \bar{f}(x_2, x_1), \bar{f}(x_3, x_2, x_1)).$$

$\triangle$
7 Symmetry and synchrony

Symmetry [5], [6], [18], [25], [31] and synchrony [2], [3], [10], [11], [27], [32], [34], [38], [49], [51], [53] have obtained much attention in the literature on coupled cell networks. They generate and explain interesting patterns, including synchronized states [30], multirhythms [19], [30], [44], [52], rotating waves [24], [32] and synchronized chaos [20], [24] and can lead to symmetry and synchrony breaking bifurcations, cf. [1], [4], [8], [9], [14], [15], [17], [22], [25], [31], [33], [43], [50]. In short, symmetry and synchrony heavily impact the dynamics and bifurcations of a network.

In this section, we relate some of the existing theory on symmetry and synchrony to the semigroup extension that we propose. More precisely, we show that the semigroup extension does not affect the symmetries or synchrony spaces of a network. This implies in particular that the symmetries and synchrony spaces of a network are also present its normal form. The semigroup extension is thus quite harmless and very natural.

To start, let us say that a permutation \( p : \{1, \ldots, N\} \to \{1, \ldots, N\} \) of the cells is a network symmetry for \( \Sigma \) if it sends the inputs of a cell to the inputs of its image. That is, if \( p \circ \sigma_j = \sigma_j \circ p \) for all \( 1 \leq j \leq n \).

The permutations with this property obviously form a group. More importantly, they are of dynamical interest because the corresponding representations

\[
\lambda_p : V^N \to V^N, \quad (x_1, \ldots, x_N) \mapsto (x_{p(1)}, \ldots, x_{p(N)})
\]

conjugate every coupled cell network map \( \gamma_f \) to itself:

\[
(\gamma_f \circ \lambda_p)_i(x) = f(\sigma_i(x_{p(1)}, \ldots, x_{p(N)})) = f(x_{p(\sigma_i(1)), \ldots, x_{p(\sigma_i(N))}}) = f(x_{\sigma_i(p(1)), \ldots, x_{\sigma_i(p(N))}}) = f(\sigma_i(x(p(1)), \ldots, x(p(N)))) = (\gamma_f(x))_{p(i)} = (\lambda_p \circ \gamma_f)_i(x).
\]

In turn, this means that when \( t \mapsto (x_1(t), \ldots, x_N(t)) \) is a solution to the differential equations \( \dot{x} = \gamma_f(x) \), then so is \( t \mapsto (x_{p(1)}(t), \ldots, x_{p(N)}(t)) \). And similarly that when \( m \mapsto (x^{(m)_1}, \ldots, x^{(m)_N}) \) is an orbit of the map \( x^{(m+1)} = \gamma_f(x^{(m)}) \), then so is \( m \mapsto (x_{p(1)}^{(m)_1}, \ldots, x_{p(N)}^{(m)_N}) \).

The following lemma states that network symmetries are trivially preserved by our semigroup extension:

**Lemma 7.1** Let \( \Sigma = \{\sigma_1, \ldots, \sigma_n\} \) be a collection of maps, not necessarily forming a semigroup, and \( p : \{1, \ldots, N\} \to \{1, \ldots, N\} \) a permutation.

Then \( p \) is a network symmetry for \( \Sigma \) if and only if it is a network symmetry for the semigroup \( \Sigma' \) generated by \( \Sigma \).

**Proof:** Elements of the semigroup \( \Sigma' \) are of the form \( \sigma_{j_1} \circ \ldots \circ \sigma_{j_l} \) for certain \( \sigma_{j_k} \in \Sigma \). But if \( p \circ \sigma_{j_k} = \sigma_{j_k} \circ p \) for \( k = 1, \ldots, l \), then also \( p \circ (\sigma_{j_1} \circ \ldots \circ \sigma_{j_l}) = (\sigma_{j_1} \circ \ldots \circ \sigma_{j_l}) \circ p \). Thus, the collection of network symmetries of \( \Sigma \) is the same as the collection of network symmetries of \( \Sigma' \). \( \square \)

Lemma 7.1 implies in particular that the composition \( \gamma_f \circ \gamma_g = \gamma_{f \circ g} \) and the Lie bracket \( [\gamma_f, \gamma_g] = \gamma_{f \circ g(\Sigma)} \) will exhibit the same network symmetries as \( \gamma_f \) and \( \gamma_g \).

Though not much more complicated, the situation is slightly more interesting for the synchronous solutions of a network. We recall that a synchrony space of a coupled cell network is an invariant subspace in which certain of the \( x_i \) (with \( 1 \leq i \leq N \)) are equal. First of all, the following result is classical, see [24], [49].

**Proposition 7.2** Let \( \Sigma = \{\sigma_1, \ldots, \sigma_n\} \) be a collection of maps, not necessarily forming a semigroup, and \( P = \{P_1, \ldots, P_r\} \) a partition of \( \{1, \ldots, N\} \). The following are equivalent:

i) For all \( 1 \leq j \leq n \) and all \( 1 \leq k_1 \leq r \) there exists a \( 1 \leq k_2 \leq r \) so that \( \sigma_j(P_{k_1}) \subset P_{k_2} \).

ii) For every \( f \in C^\infty(V^N, V) \) the subspace

\[
\text{Sym}_P := \{ x \in V^N \mid x_{i_1} = x_{i_2} \text{ when } i_1 \text{ and } i_2 \text{ are in the same element of } P \}
\]

is an invariant submanifold for the dynamics of \( \gamma_f \).
Proof: The subspace $\text{Syn}_P$ is invariant under the flow of the differential equation $\dot{x} = \gamma_f(x)$ if and only if the vector field $\gamma_f$ is tangent to $\text{Syn}_P$. Similarly, $\text{Syn}_P$ is invariant under the map $x^{(m+1)} = \gamma_f(x^{(m)})$ if and only if $\gamma_f$ sends $\text{Syn}_P$ to itself. Both properties just mean that for all $x \in \text{Syn}_P$ it holds that

$$f(x_{\sigma_1(i_1)}, \ldots, x_{\sigma_n(i_1)}) = f(x_{\sigma_1(i_2)}, \ldots, x_{\sigma_n(i_2)})$$

for all $i_1, i_2$ in the same element of $P$.

The latter statement holds for all $f \in C^\infty(V^n, V)$ if and only if for all $x \in \text{Syn}_P$,

$$x_{\sigma_j(i)} = x_{\gamma_j(i)}$$

for all $1 \leq j \leq n$ and all $i_1, i_2$ in the same element of $P$.

It is not hard to see that this is true precisely when all $\sigma_j \in \Sigma$ map the elements of $P$ into elements of $P$. \hfill $\Box$

A partition $P$ of $\{1, \ldots, N\}$ with property $i)$ is sometimes called a balanced partition or balanced coloring and a subspace $\text{Syn}_P$ satisfying property $ii)$ a (robust) synchrony space.

The following result says that the synchrony spaces of a network do not change if one extends the network architecture to a semigroup:

**Lemma 7.3** Let $\Sigma = \{\sigma_1, \ldots, \sigma_n\}$ be a collection of maps, not necessarily forming a semigroup, and $P = \{P_1, \ldots, P_r\}$ a partition of $\{1, \ldots, N\}$.

Then $\text{Syn}_P$ is a (robust) synchrony space for $\Sigma$ if and only if it is a (robust) synchrony space for the semigroup $\Sigma'$ generated by $\Sigma$.

Proof: Elements of the semigroup $\Sigma'$ are of the form $\sigma_{i_1} \circ \ldots \circ \sigma_{i_k}$ for certain $\sigma_{j_k} \in \Sigma$. This implies that the elements of $\Sigma$ send the elements of $P$ inside elements of $P$ if and only if the elements of $\Sigma'$ do. In other words: that the collection of balanced partitions of $\Sigma$ and of $\Sigma'$ are the same. The result now follows from Proposition 7.2. \hfill $\Box$

Lemma 7.3 implies in particular that the composition $\gamma_f \circ \gamma_g = \gamma_{f \circ \gamma_l}$ and the Lie bracket $[\gamma_f, \gamma_g] = \gamma_{[f, g]_{\Sigma'}}$ will exhibit the same synchrony spaces as $\gamma_f$ and $\gamma_g$.

We conclude this section with the following simple but important observation:

**Corollary 7.4** Let $\Sigma = \{\sigma_1, \ldots, \sigma_n\}$ be a collection of maps, not necessarily forming a semigroup, and $\gamma_f$ a coupled cell network vector field subject to $\Sigma$.

Then a local normal form $\gamma_{\gamma_f}$ of $\gamma_f$ has the same network symmetries and the same synchrony spaces as $\gamma_f$.

Proof: $\gamma_{\gamma_f}$ is a coupled cell network with respect to the semigroup $\Sigma'$ generated by $\Sigma$. Thus, the result follows from Lemma 7.1 and Lemma 7.3. \hfill $\Box$

**Example 7.5** Again, let $N = 3$ and let $\sigma_1, \sigma_2, \sigma_3$ be defined as in Examples 2.3 and 2.4. Recall that a coupled cell network differential equation subject to $\{\sigma_1, \sigma_2\}$ is of the form

$$\dot{x}_1 = f(x_1, x_1), \dot{x}_2 = f(x_2, x_1), \dot{x}_3 = f(x_3, x_2).$$

These equations do not have any network symmetries, but they do admit the nontrivial balanced partitions

$$\{1, 2\} \cup \{3\} \text{ and } \{1, 2, 3\}. $$

The corresponding invariant synchrony spaces $\{x_1 = x_2\}$ and $\{x_1 = x_2 = x_3\}$ are preserved in the normal form, because the latter is a coupled cell network subject to $\{\sigma_1, \sigma_2, \sigma_3\}$. \hfill $\triangle$

8 Input symmetries

As mentioned in Remark 2.2, the admissible maps and vector fields of a network system can be determined from its groupoid of input equivalences, see [35], [38] and [51]. This groupoid consists of all possible ways to identify the inputs of cell $i$ with the inputs of cell $j$, for all the cells $1 \leq i, j \leq N$.

In case the network is homogeneous, the groupoid of the network is the obvious one that identifies the $k$-th input of cell $i$ with the $k$-th input of cell $j$ for all $1 \leq k \leq n$. No inputs can therefore be interchanged and, putting it equivalently, there exists only one way to identify
the inputs of one cell among themselves. The so-called vertex group of the groupoid therefore consists of only one element.

On the other hand, many coupled cell networks that appear in the mathematical literature display a nontrivial network groupoid. This case arises when the response function \( f \) is assumed invariant under the permutation of some of its inputs. Such an input symmetry can be dynamically relevant because it may give rise to nontrivial robust synchrony spaces.

In this section, we point out one condition under which an input symmetry can be preserved in the normal form of \( f \). Although rather intuitive, this condition appears far from optimal. As a consequence, this section is not important for the remainder of this paper and can be skipped at first reading.

Concretely, an input symmetry is reflected by a permutation of inputs \( q: \{1, \ldots, n\} \rightarrow \{1, \ldots, n\} \) with the property that

\[
f(x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)}) = f(x_{\sigma_q(1)(i)}, \ldots, x_{\sigma_q(n)(i)}) \quad \text{for all } x \in V^N \text{ and all } 1 \leq i \leq N. \tag{8.19}
\]

In other words, we require that \( f \circ \lambda_q \circ \pi_i = f \circ \pi_i \) for all \( 1 \leq i \leq N \). Perhaps remarkably, the permutations \( q \) for which (8.19) holds in general need not form a group.

On the other hand, one can also consider only those input symmetries that give rise to a network symmetry. This means that along with the permutation \( q \) of the inputs \( \{1, \ldots, n\} \) there exists a permutation \( p \) of the cells \( \{1, \ldots, N\} \) that sends the \( j \)-th input of each cell to the \( q(j) \)-th input of its image, i.e. that

\[
p \circ \sigma_j = \sigma_{q(j)} \circ p \quad \text{for all } 1 \leq j \leq n. \tag{8.20}
\]

**Definition 8.1** We call a permutation \( q \) of the inputs for which there exists a permutation \( p \) of the cells so that (8.19) and (8.20) hold a **dynamical input symmetry** of \( \gamma_f \).

**Lemma 8.2** The dynamical input symmetries of \( \gamma_f \) form a group.

Before we prove Lemma 8.2, let us translate property (8.20) as follows:

**Proposition 8.3** When (8.20) holds, then

\[\pi_1 \circ \lambda_p = \lambda_q \circ \pi_{p(i)} \quad \text{for all } 1 \leq i \leq N.\]

**Proof:** This follows from a little computation:

\[
(\pi_1 \circ \lambda_p)(x) = \pi_1(x_{p(1)}, \ldots, x_{p(N)}) = (x_{p(\sigma_1(1))}, \ldots, x_{p(\sigma_n(1))}) = (x_{\sigma_1(\sigma_1(1))}, \ldots, x_{\sigma_1(\sigma_n(1))}) = (\lambda_q \circ \pi_{p(1)})(x).
\]

**Proof** (of Lemma 8.2): Assume that \( p_1 \circ \sigma_j = \sigma_{q_1(j)} \circ p_1 \) and \( p_2 \circ \sigma_j = \sigma_{q_2(j)} \circ p_2 \) for all \( 1 \leq j \leq n \) and that \( f \circ \lambda_{q_1} \circ \pi_i = f \circ \pi_i \) and \( f \circ \lambda_{q_2} \circ \pi_i = f \circ \pi_i \) for all \( 1 \leq i \leq N \).

Then it follows that

\[
(p_1 \circ p_2) \circ \sigma_j = p_1 \circ \sigma_{q_2(j)} \circ p_2 = \sigma_{(q_1 \circ q_2)(j)} \circ (p_1 \circ p_2)
\]

and, using (8.21), that

\[
f \circ \lambda_{q_1 \circ q_2} \circ \pi_i = f \circ \lambda_{q_2} \circ \lambda_{q_1} \circ \pi_i = f \circ \lambda_{q_2} \circ \pi_{p_1^{-1}(i)} \circ \lambda_{p_1} = f \circ \pi_{p_1^{-1}(i)} \circ \lambda_{p_1} = f \circ \lambda_{q_1} \circ \pi_i = f \circ \pi_i.
\]

This proves that the dynamical input symmetries form a group.

Not surprisingly, dynamical input symmetries give rise to a symmetry of the dynamics of the network:

**Proposition 8.4** Let \( p \) be a permutation of \( \{1, \ldots, N\} \) and \( q \) a permutation of \( \{1, \ldots, n\} \). Assume that \( p \circ \sigma_j = \sigma_{q(j)} \circ p \) for all \( 1 \leq j \leq n \) and that (8.19) holds. Then

\[
\gamma_f \circ \lambda_p = \lambda_{p} \circ \gamma_f.
\]
Proof: From (8.21) and our assumption that \( f \circ \lambda_q = f \) on every \( \text{im} \pi_i \), it follows that
\[
(\gamma_f \circ \lambda_p)_i = f \circ \pi_i \circ \lambda_p = f \circ \lambda_q \circ \pi_{p(i)} = f \circ \pi_{p(i)} = (\gamma_f)_{p(i)} = (\lambda_p \circ \gamma_f)_i.
\]
\[\square\]

The relevance of dynamical input symmetries for normal forms will be explained below. We first show that dynamical input symmetries survive the semigroup extension.

**Lemma 8.5** Let \( \Sigma = \{\sigma_1, \ldots, \sigma_n\} \) be a collection of maps, not necessarily forming a semigroup and let \( q : \{1, \ldots, n\} \rightarrow \{1, \ldots, n\} \) be a dynamical input symmetry for \( \Sigma \).

Then the latter extends to a unique dynamical input symmetry
\[
q' : \{1, \ldots, n, n+1, \ldots, n'\} \rightarrow \{1, \ldots, n, n+1, \ldots, n'\}
\]
for the semigroup \( \Sigma' = \{\sigma_1, \ldots, \sigma_n, \sigma_{n+1}, \ldots, \sigma_{n'}\} \) generated by \( \Sigma \).

**Proof:** Recall that elements of \( \Sigma' \) are of the form \( \sigma_{j_1} \circ \ldots \circ \sigma_{j_n} \) for certain \( \sigma_{j_k} \in \Sigma \). Assume now that \( p \circ \sigma_{j_1} = \sigma_{q(j_1)} \circ p \) and \( p \circ \sigma_{j_2} = \sigma_{q(j_2)} \circ p \). Then it follows that
\[
p \circ \sigma_{j_1} (j_2) = p \circ (\sigma_{j_1} \circ \sigma_{j_2}) = \sigma_{q(j_1)} \circ \sigma_{q(j_2)} = \sigma_{q(j_1) \circ q(j_2)} \circ p \circ \sigma_{j_2} = \sigma_{q(j_1) \circ q(j_2)} \circ p.
\]
This means that if an extension \( q' : \{1, \ldots, n, n+1, \ldots, n'\} \rightarrow \{1, \ldots, n, n+1, \ldots, n'\} \) exists, then it must be unique and satisfy
\[
q'(\sigma_{j_1}(j_2)) = \sigma_{q(j_1)}(q(j_2)).
\]
If now \( p \circ \sigma_{j_1} = \sigma_{q(j_1)} \circ p \) and \( p \circ \sigma_{j_2} = \sigma_p(j_2) \circ p \) and \( \tilde{\sigma}_{j_1}(j_2) = \sigma_{j_1}(j_2) = \tilde{\sigma}_{j_1}(j_2) \), then actually
\[
\sigma_{q(j_1)}(q(j_2)) = \tilde{\sigma}_{q(j_1)}(q(j_2)) \] if \( \tilde{\sigma}_{j_1}(j_2) = \sigma_{j_1}(j_2) \) and hence that \( q' \) is well-defined.
\[\square\]

The following result explains that network symmetries are preserved under taking compositions and Lie brackets:

**Theorem 8.6** Let \( \Sigma \) be a semigroup and assume \( p \circ \sigma_j = \sigma_{q(j)} \circ p \) for all \( 1 \leq j \leq n \). Then
\[
(f \circ g \circ \lambda_q = (f \circ \lambda_q) \circ \Sigma \circ (g \circ \lambda_q)) \text{ on every } \text{im} \pi_i,
\]
\[
[f, g]_{\Sigma} \circ \lambda_q = [f \circ \lambda_q, g \circ \lambda_q]_{\Sigma} \text{ on every } \text{im} \pi_i. \tag{8.22}
\]

**Proof:** With slight abuse of notation, we write
\[
(f \circ g) \circ \lambda_q \circ \pi_{p(i)} = f(\ldots, f(\ldots, f(g(A_{\sigma_j} \circ \lambda_q \circ \pi_{p(i)}), \ldots), \ldots) = f(\ldots, f(\ldots, f(g(\pi_{\sigma_j(i)} \circ \lambda_q), \ldots), \ldots) = f(\ldots, f(\ldots, f(g \circ \lambda_q(\pi_{\sigma_j(i)}(p(i))), \ldots), \ldots) = f(\ldots, f(\ldots, f(g \circ \lambda_q(\pi_{\sigma_j(i)}(p(i))), \ldots), \ldots) = f(\ldots, f(\ldots, f(g \circ \lambda_q(\pi_{\sigma_j(i)}(p(i))), \ldots), \ldots) = f(g \circ \lambda_q(\pi_{\sigma_j(i)}(p(i))), \ldots) = (f \circ \lambda_q) \circ \Sigma(g \circ \lambda_q) \circ \pi_{p(i)}.
\]
This proves that \( (f \circ g) \circ \lambda_q = (f \circ \lambda_q) \circ \Sigma(g \circ \lambda_q) \) on each \( \text{im} \pi_{p(i)} \) and hence, because \( p \) is invertible, on each \( \text{im} \pi_i \). The proof for the Lie bracket is similar.
\[\square\]

Finally, we conclude

**Corollary 8.7** Let \( \Sigma = \{\sigma_1, \ldots, \sigma_n\} \) be a collection of maps, not necessarily forming a semigroup, and \( \gamma_f \) a coupled cell network vector field subject to \( \Sigma \).

Then the local normal form \( \gamma_f \) of \( \gamma_f \) can be chosen to have the same dynamical input symmetries as \( \gamma_f \).

**Proof:** Let \( G \) denote the group of dynamical input symmetries of \( f = f_0 + f_1 + f_2 + \ldots \) and let us define the set of \( G \)-invariant functions
\[
C^\infty_G(V^n, V) := \{g \in C^\infty(V^n, V) \mid g \circ \lambda_q \circ \pi_i = g \circ \pi_i \text{ for all } q \in G \text{ and all } 1 \leq i \leq N\}.
\]
Theorem 8.6 implies that if \( g, h \in C^\infty_G(V^n, V) \), then also
\[
[g, h] \in C^\infty_G(V^n, V).
\]
The fact that \( f \in C^\infty_G(V^n, V) \) moreover implies that
\[
f_k \in P^k_G := \{ g_k \in P^k | g_k \circ \lambda_q \circ \pi_i = g_k \circ \pi_i \text{ for all } q \in G \text{ and all } 1 \leq i \leq N \}
\] is a \( G \)-invariant polynomial of degree \( k+1 \).

It clearly holds that \([P^k_G, P^l_G] \subseteq P^{k+l}_G\). As a consequence, we can repeat the proof of Theorem 6.1 by replacing \( P^k \) by \( P^k_G \) and choosing the normal form spaces \( N^k_G \subseteq P^k_G \) so that
\[
\text{im ad}_{f_0}|_{P^k_G} \oplus N^k_G = P^k_G.
\]
This produces a normal form \( \mathcal{F} \in C^\infty_G(V^n, V) \). \( \Box \)

**Example 8.8** Consider the class of differential equations of the form
\[
\begin{align*}
\dot{x}_1 &= f(x_1, x_2, x_2, x_1), \\
\dot{x}_2 &= f(x_1, x_2, x_1, x_2).
\end{align*}
\]
These differential equations have a semigroup coupled cell network structure with \( N = 2 \) and \( n = 4 \), see Figure 3.

![Figure 3: The network with \( N = 2 \) and \( n = 4 \).](image)

The semigroup \( \Sigma \) in this case is the full non-Abelian semigroup of maps on 2 symbols. In other words, \( \Sigma = \{ \sigma_1, \sigma_2, \sigma_3, \sigma_4 \} \), where
\[
\begin{align*}
\sigma_1(1) &= 1, \sigma_1(2) = 1, \sigma_2(1) = 2, \sigma_2(2) = 2, \sigma_3(1) = 2, \sigma_3(2) = 1, \sigma_4(1) = 1, \sigma_4(2) = 2.
\end{align*}
\]
There is only one nontrivial permutation \( p : \{1, 2\} \to \{1, 2\} \) of the cells, which is defined by \( p(1) := 2 \) and \( p(2) := 1 \). It is easily checked that
\[
p \circ \sigma_1 = \sigma_2 \circ p, \quad p \circ \sigma_2 = \sigma_1 \circ p, \quad p \circ \sigma_3 = \sigma_3 \circ p, \quad p \circ \sigma_4 = \sigma_4 \circ p.
\]
In other words, \( p \circ \sigma_j = \sigma_{q(j)} \circ p \) if we let \( q : \{1, 2, 3, 4\} \to \{1, 2, 3, 4\} \) be defined by
\[
q(1) = 2, q(2) = 1, q(3) = 3, q(4) = 4.
\]
Thus, the (in this case identical) invariances
\[
\begin{align*}
f(x_1, x_2, x_2, x_1) &= f(x_2, x_1, x_2, x_1), \\
f(x_1, x_2, x_1, x_2) &= f(x_2, x_1, x_1, x_2)
\end{align*}
\]
can be preserved in the normal form of \( f \). These are precisely the invariances that make \( \lambda_p : (x_1, x_2) \mapsto (x_2, x_1) \) a symmetry of the differential equations. \( \triangle \)
9 SN-decomposition

We recall from the previous sections that when $f_0 \in L(V^n, V)$, then $\text{ad}_{f_0}^{\Sigma} : P^k \to P^k$. The operators $\text{ad}_{f_0}^{\Sigma} |_{p_k}$ are called “homological operators" and they play an important role in normal form theory. This is first of all because the normal form spaces $N^k \subset P^k$ of Theorem 6.1 must be chosen complementary to their images, and secondly because in computing a normal form one needs to “invert" them when solving the homological equations $\text{ad}_{f_0}^{\Sigma}(g_k) - h_k \in N^k$, see the proof of Theorem 6.1.

For this reason, it is convenient to have at one’s disposal the “SN-decompositions” (also called “Jordan-Chevalley decompositions”) of the homological operators $[\gamma]$ of the homological operators $[\Sigma]$. We recall that, since $P^k$ is finite-dimensional, the map $\text{ad}_{f_0}^{\Sigma} |_{p_k}$ admits a unique SN-decomposition

$$\text{ad}_{f_0}^{\Sigma} |_{p_k} = (\text{ad}_{f_0}^{\Sigma} |_{p_k})^S + (\text{ad}_{f_0}^{\Sigma} |_{p_k})^N$$

(9.23)

in which the map $(\text{ad}_{f_0}^{\Sigma} |_{p_k})^S$ is semisimple, the map $(\text{ad}_{f_0}^{\Sigma} |_{p_k})^N$ is nilpotent and the two maps $(\text{ad}_{f_0}^{\Sigma} |_{p_k})^S$ and $(\text{ad}_{f_0}^{\Sigma} |_{p_k})^N$ commute. The aim of this somewhat technical section is to describe this SN-decomposition in an as simple as possible way.

We start with the following remark: when $f_0 \in L(V^n, V)$, then $\gamma f_0 \in L(V^N, V^N)$ and hence the latter has an SN-decomposition

$$\gamma f_0 = \gamma^S f_0 + \gamma^N f_0$$

for certain $\gamma^S f_0, \gamma^N f_0 \in L(V^N, V^N)$. We are going to relate this SN-decomposition of $\gamma f_0$ to the SN-decomposition of $\text{ad}_{f_0}^{\Sigma} |_{p_k}$. In order to do so, it is first of all important to observe that $\gamma f_0$ and $\gamma^N f_0$ are themselves coupled cell network maps:

**Lemma 9.1** For every $f_0 \in L(V^n, V)$ there exist $f_0^S, f_0^N \in L(V^n, V)$ with $f_0 = f_0^S + f_0^N$ and

$$\gamma^S f_0 = \gamma f_0^S \text{ and } \gamma^N f_0 = \gamma f_0^N.$$  

**Proof:** We recall - see for instance [39], pp. 17 - that both the semisimple part $\gamma^S f_0$ and the nilpotent part $\gamma^N f_0$ are polynomial functions of $\gamma f_0$. More precisely, $\gamma^S f_0 = p(\gamma f_0)$ and $\gamma^N f_0 = p(\gamma f_0)$, where

$$p(\gamma) = a_0 I + a_1 \gamma + \ldots + a_d \gamma^d$$

is a polynomial with coefficients $a_0, \ldots, a_d \in \mathbb{C}$.

Theorem 4.2 then implies that $\gamma^S f_0 = p(\gamma f_0) = \gamma f_0^S$ for $f_0^S \in L(V^n, V)$ defined as

$$f_0^S = p(f_0) = a_0 + a_1 f_0 + a_2 f_0 \circ f_0 + \ldots + a_d f_0 \circ \ldots \circ f_0,$$  

(9.24)

where $i : V^n \to V$ satisfies $\gamma_i = \text{id}_{V^N}$. By Lemma 4.3 this $f_0^S$ is well-defined. Similarly, $\gamma f_0 = f_0^N$ for a well-defined $f_0^N = f_0 - p(f_0) \in L(V^n, V)$. Clearly, $f_0 = f_0^S + f_0^N$.

One would maybe like to claim that

$$\text{ad}_{f_0}^{\Sigma} |_{p_k} = \text{ad}_{f_0^S}^{\Sigma} |_{p_k} + \text{ad}_{f_0^N}^{\Sigma} |_{p_k}$$

is an SN-decomposition, but this need not be true. The problem is that the $f_0^S$ and $f_0^N$ of Lemma 9.1 (for which $\gamma f_0 = \gamma f_0^S$ and $\gamma f_0 = \gamma f_0^N$) need not be unique - while SN-decompositions are. In fact, two functions $f, g \in C^\infty(V^n, V)$ generate the same network map (in the sense that $\gamma f = \gamma g$) if and only if $f - g \in \ker \gamma = \{h \mid h|_{V^n} = 0\} \subset C^\infty(V^n, V)$. So whenever $\ker \gamma$ is nontrivial, then $f_0^S$ and $f_0^N$ are not uniquely determined.

This ambiguity can be removed though, by observing that when $f, g \in C^\infty(V^n, V)$ and $f \in \ker \gamma$, then $\gamma f \circ g|_{V^n} = \gamma f |_{V^n} = \{h \mid h|_{V^n} = 0\} = \ker \gamma$. Hence also $[f, g]|_{V^n} \in \ker \gamma$. This means $\ker \gamma \subset C^\infty(V^n, V)$ is a Lie algebra ideal. In particular it holds for every $f \in C^\infty(V^n, V)$ that the adjoint map $\text{ad}_{f}^{\Sigma} : C^\infty(V^n, V) \to C^\infty(V^n, V)$ sends $\ker \gamma$ to $\ker \gamma$ and hence that $\text{ad}_{f}^{\Sigma}$ descends to a well-defined map

$$\text{ad}_{f}^{\Sigma} : C^\infty(V^n, V) / \ker \gamma \to C^\infty(V^n, V) / \ker \gamma.$$  

With this in mind, we can formulate the main result of this section, that explains how the decomposition $f_0 = f_0^S + f_0^N$ determines the SN-decomposition of $\text{ad}_{f_0}^{\Sigma} |_{p_k}$:
Theorem 9.2 For every $k = 0, 1, 2, \ldots$ the maps

$$
(\text{ad}^S_{f_{0}}|_{p_k})^S \text{ and } \text{ad}^S_{f_{0}}|_{p_k} \text{ respectively (ad}^S_{f_{0}}|_{p_k})^N \text{ and } \text{ad}^S_{f_{0}}|_{p_k}
$$
descend to the same map on $P^k/\ker \gamma$.

Proof: We start by repeating that $\text{ad}^S_{f_{0}}$ maps $\ker \gamma$ into itself and hence descends to a well-defined map on $C^\infty(V^n, V)/\ker \gamma$ that moreover sends $P^k/\ker \gamma$ into itself. More interestingly, since $(\text{ad}^S_{f_{0}}|_{p_k})^S$ and $(\text{ad}^S_{f_{0}}|_{p_k})^N$ are polynomial functions of $\text{ad}^S_{f_{0}}|_{p_k}$, also these latter maps send $\ker \gamma$ into itself and thus descend to $P^k/\ker \gamma$.

For the actual proof of the theorem, we define for $k \geq 0$ the vector spaces

$$
\mathcal{P}^k := \{\text{homogeneous polynomial vector fields on } V^n \text{ of degree } k+1\}.
$$

It is clear that $\gamma : P^k/\ker \gamma \to \mathcal{P}^k$ is an injective linear map. Moreover, the computation

$$
(\gamma \circ \text{ad}^S_{f_{0}})(g_k) = [\gamma_{f_{0}, g_{0}}, \gamma_{g_{0}}] = (\text{ad}^S_{f_{0}, g_{0}} \circ \gamma)(g_{0})
$$

reveals that the maps $\text{ad}^S_{f_{0}} : P^k/\ker \gamma \to P^k/\ker \gamma$ and $\gamma_{f_{0}, g_{0}} : \mathcal{P}^k \to \mathcal{P}^k$ are conjugate by the map $\gamma : P^k/\ker \gamma \to \mathcal{P}^k$. Similarly, $\text{ad}^S_{f_{0}}$ is conjugate to $\gamma_{f_{0}}$ and $\text{ad}^S_{f_{0}}$ is conjugate to $\gamma_{f_{0}}$. Now we recall the well-known fact that the SN-decomposition of $\text{ad}^S_{f_{0}}|_{p_k}$ is

$$
\text{ad}^S_{f_{0}}|_{p_k} = \left(\text{ad}^S_{f_{0}}|_{p_k}\right)^S + \left(\text{ad}^S_{f_{0}}|_{p_k}\right)^N = \text{ad}^S_{f_{0}}|_{p_k} + \text{ad}^N_{f_{0}}|_{p_k} = \gamma_{f_{0}} + \gamma_{f_{0}}|_{p_k}.
$$

Because $\gamma$ is injective, we have thus proved that

$$
\text{ad}^S_{f_{0}}|_{p_k} = \left(\text{ad}^S_{f_{0}}|_{p_k}\right)^S + \left(\text{ad}^S_{f_{0}}|_{p_k}\right)^N : P^k/\ker \gamma \to P^k/\ker \gamma
$$

is the SN-decomposition of the quotient map. Because the SN-decomposition of the quotient is the quotient of the SN-decomposition, this proves the theorem.

Because the elements of $\ker \gamma$ are dynamically completely irrelevant, for all practical purposes we can think of Theorem 9.2 as saying that

$$
\text{ad}^S_{f_{0}}|_{p_k} = \left(\text{ad}^S_{f_{0}}|_{p_k}\right)^S + \left(\text{ad}^S_{f_{0}}|_{p_k}\right)^N \text{ is the SN-decomposition of } \text{ad}^S_{f_{0}}|_{p_k}.
$$

This is very convenient, because it means that one can determine the SN-decompositions of all operators $\text{ad}^S_{f_{0}}|_{p_k}$ simultaneously by simply determining the splitting $f_{0} = f_{0}^S + f_{0}^N$.

Example 9.3 The map $\gamma : C^\infty(V^n, V) \to C^\infty(V^n, V)$ fails to be injective as soon as $\bigcup_{i=1}^{n} \text{im } \pi_{i} \neq V^n$, because $\gamma_{f} = 0$ already when $f$ vanishes on every im $\pi_{i}$. This is the reason for the somewhat difficult formulation of Theorem 9.2.

To illustrate this phenomenon, we refer to Example 8.8 in which

$$
\pi_{1}(x_{1}, x_{2}) = (x_{1}, x_{2}, x_{2}, x_{1}) \text{ and } \pi_{2}(x_{1}, x_{2}) = (x_{1}, x_{2}, x_{1}, x_{2}),
$$

so that in particular im $\pi_{1} \cup \text{im } \pi_{2} \neq V^4$. In this case it is even true that im $\pi_{1} + \text{im } \pi_{2} \neq V^4$, because both images are 2-dimensional and intersect in a 1-dimensional space. For example, $X_{1} + X_{2} - X_{3} - X_{4} \in \ker \gamma \cap L(V^n, V)$. 

We conclude this section with the following dynamical implication of Theorem 9.2:

Corollary 9.4 Let $0 \leq r < \infty$. Then it can be arranged that the normal form $\bar{f} = f_{0} + f_{1} + \ldots \in C^\infty(V^n, V)$ of an $f = f_{0} + f_{1} + \ldots \in C^\infty(V^n, V)$ has the special property that the truncated normal form coupled cell map/vector field $\gamma_{f_{0} + f_{1} + \ldots + f_{r}}$ commutes with the continuous family of maps

$$
t \mapsto e^{t \gamma_{f_{0}}^S}.
$$
Proof: Recall that each of the normal form spaces $N^k \subset P^k$ of Theorem 6.1 is required to have the property that $N^k \oplus \text{im} \ (\text{ad}_{f_0}^S \big|_{P^k}) = P^k$. It is not hard to see that one can choose such $N^k \subset \ker (\text{ad}_{f_0}^S \big|_{P^k})^S$. Indeed, whenever

\[ N^k \subset \ker (\text{ad}_{f_0}^S \big|_{P^k})^S \text{ is complementary to } \text{im} (\text{ad}_{f_0}^S \big|_{P^k}) \cap \ker (\text{ad}_{f_0}^S \big|_{P^k})^S, \]

then $N^k \oplus \text{im} \ (\text{ad}_{f_0}^S \big|_{P^k}) = P^k$.

Thus, let us choose $N^k \subset \ker (\text{ad}_{f_0}^S \big|_{P^k})^S$. The fact that $(\text{ad}_{f_0}^S \big|_{P^k})^S$ and $(\text{ad}_{f_0}^S \big|_{P^k})^S$ descend to the same map on $P^k / \ker \gamma$ then implies in particular that $\text{ad}_{f_0}^S (N^k) \subset \ker \gamma$.

Let now $T = f_0 + T_1 + T_2 + \ldots$ be any normal form of $f$ of order $r$ with respect to the $N^k$, meaning that $T_k \in N^k$ for all $1 \leq k \leq r$. Such a normal form exists by Theorem 6.1. Then it holds that $\text{ad}_{f_0}^S (T_k) \in \ker \gamma$ and in view of Theorem 4.2 we therefore have

\[ [\gamma I^S, \gamma f_0 + T_1 + \ldots + T_r] = \gamma [I^S, f_0 + T_1 + \ldots + T_r] = 0. \]

Hence, $\frac{\partial}{\partial t} \big|_{t=0} (e^{t I^S}) \gamma f_0 + T_1 + \ldots + T_r = 0$ and thus the truncated normal form commutes with the flow $t \mapsto e^{t I^S}$ of the coupled cell network vector field $\gamma I^S$.

The continuous family

\[ t \mapsto e^{t I^S} \]

of transformations of $V^N$ is called a normal form symmetry. This symmetry is sometimes used to characterize vector fields that are in normal form. It also plays an important role in finding periodic solutions near equilibria of the vector field $\gamma f$, using for example the method of Lyapunov-Schmidt reduction [12], [23], [28], [33], [37].

10 A fundamental semigroup network

As a byproduct of Theorem 4.1, and perhaps as a curiosity, we will show in this section that the dynamics of $\gamma f$ on $V^N$ is conjugate to the dynamics of a certain network $\Gamma f$ on $V^N$. We will argue that $\Gamma f$ acts as a “fundamental network” for $\gamma f$.

We recall that if $\Sigma = \{\sigma_1, \ldots, \sigma_n\}$ is a semigroup, then every $\sigma_j \in \Sigma$ induces a map

\[ \overline{\sigma}_j : \{1, \ldots, n\} \rightarrow \{1, \ldots, n\} \text{ via the formula } \sigma_{\overline{\sigma}_j(k)} = \sigma_j \circ \sigma_k. \]

We saw that $\sigma_{\overline{\sigma}_1} \circ \sigma_{\overline{\sigma}_2} = \overline{\sigma}_1 \circ \overline{\sigma}_2$ and hence the collection $\overline{\Sigma} := \{\overline{\sigma}_1, \ldots, \overline{\sigma}_n\}$ is closed under composition. One can now study coupled cell networks subject to $\overline{\Sigma}$. They have the form

\[ \Gamma f : V^N \rightarrow V^N \text{ with } (\Gamma f)_{\sigma_j} (X) := f(X_{\overline{\sigma}_1(j)}, \ldots, X_{\overline{\sigma}_n(j)}) = f(A_{\sigma_j} X) \text{ for all } 1 \leq j \leq n. \]

The following theorem demonstrates that $\gamma f$ and $\Gamma f$ are dynamically related:

**Theorem 10.1** All maps $\pi_i : V^N \rightarrow V^N$ conjugate $\gamma f$ to $\Gamma f$, that is

\[ \Gamma f \circ \pi_i = \pi_i \circ \gamma f \text{ for all } 1 \leq i \leq N. \]

**Proof:** For $x \in V^N$ we have that

\[ (\Gamma f \circ \pi_i)_j (x) = f((A_{\sigma_j} \circ \pi_i)(x)) = f(\pi_{\sigma_j(i)}(x)) = (\gamma f(x))_{\sigma_j(i)} = (\pi_i \circ \gamma f)_j (x). \]

\[ \square \]

Theorem 10.1 implies that every $\pi_i$ sends integral curves of $\gamma f$ to integral curves of $\Gamma f$ and discrete-time orbits of $\gamma f$ to discrete-time orbits of $\Gamma f$.

In addition, the dynamics of $\gamma f$ can be reconstructed from the dynamics of $\Gamma f$. More precisely, when $X_{(i)}(t)$ are integral curves of $\Gamma f$ with $X_{(i)}(0) = \pi_i(x(0))$, then an integral curve $x(t)$ of $\gamma f$ can simply be obtained by integration of the equations

\[ \dot{x}_i(t) = f(X_{(i)}(t)) \text{ for } 1 \leq i \leq N. \]
Similarly, if $X_{(i)}^{(m)}$ are discrete-time orbits of $\Gamma_f$ with $X_{(i)}^{(m)}(0) = \pi_i(x(0))$, then $x_{(i)}^{(m+1)} := f(X_{(i)}^{(m)})$ defines a discrete-time orbit of $\gamma_f$.

The transition from $\gamma_f$ to $\Gamma_f$ is thus reminiscent of the symmetry reduction of an equivariant dynamical system: the dynamics of $\gamma_f$ descends to the dynamics of $\Gamma_f$ and the dynamics of $\gamma_f$ can be reconstructed from that of $\Gamma_f$ by means of integration. Nevertheless, $n$ can of course be both smaller and larger than $N$. In the latter case, the dynamics of $\Gamma_f$ may be much richer than that of $\gamma_f$ and it is confusing to speak of reduction. In either case, $\Gamma_f$ captures all the dynamics of $\gamma_f$.

**Example 10.2** Again, let $N = 3$ and let $\sigma_1, \sigma_2, \sigma_3$ be defined as in Example 2.4. Recall

$$
A_{\sigma_1}(X_1, X_2, X_3) = (X_1, X_2, X_3),
A_{\sigma_2}(X_1, X_2, X_3) = (X_2, X_3, X_3),
A_{\sigma_3}(X_1, X_2, X_3) = (X_3, X_3, X_3).
$$

This means that the network map $\Gamma_f$ is given by

$$
\Gamma_f(X_1, X_2, X_3) = (f(X_1, X_2, X_3), f(X_2, X_3, X_3), f(X_3, X_3, X_3)).
$$

In this example, the conjugacies from $\gamma_f$ to $\Gamma_f$ are

$$
\begin{align*}
\pi_1 : (x_1, x_2, x_3) &\mapsto (X_1, X_2, X_3) := (x_1, x_1, x_1), \\
\pi_2 : (x_1, x_2, x_3) &\mapsto (X_1, X_2, X_3) := (x_2, x_1, x_1), \\
\pi_3 : (x_1, x_2, x_3) &\mapsto (X_1, X_2, X_3) := (x_3, x_2, x_1).
\end{align*}
$$

The conjugacy $\pi_3$ is bijective, which explains that Figures 1 and 4 are isomorphic.

\begin{figure}[h]
\centering
\includegraphics[width=0.7\textwidth]{network.png}
\caption{The fundamental network of our three-cell feedforward network.}
\end{figure}

**Example 10.3** Recall Example 8.8 that features a semigroup with 4 elements. To determine the maps $A_{\sigma}$, we compute the multiplication table of this semigroup:

<table>
<thead>
<tr>
<th></th>
<th>$\sigma_1$</th>
<th>$\sigma_2$</th>
<th>$\sigma_3$</th>
<th>$\sigma_4$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\sigma_1$</td>
<td>$\sigma_1$</td>
<td>$\sigma_1$</td>
<td>$\sigma_1$</td>
<td>$\sigma_1$</td>
</tr>
<tr>
<td>$\sigma_2$</td>
<td>$\sigma_2$</td>
<td>$\sigma_2$</td>
<td>$\sigma_2$</td>
<td>$\sigma_2$</td>
</tr>
<tr>
<td>$\sigma_3$</td>
<td>$\sigma_1$</td>
<td>$\sigma_4$</td>
<td>$\sigma_4$</td>
<td>$\sigma_4$</td>
</tr>
<tr>
<td>$\sigma_4$</td>
<td>$\sigma_1$</td>
<td>$\sigma_2$</td>
<td>$\sigma_3$</td>
<td>$\sigma_4$</td>
</tr>
</tbody>
</table>

This implies that

$$
\begin{align*}
A_{\sigma_1}(X_1, X_2, X_3, X_4) &= (X_1, X_2, X_2, X_1), \\
A_{\sigma_2}(X_1, X_2, X_3, X_4) &= (X_1, X_2, X_1, X_2), \\
A_{\sigma_3}(X_1, X_2, X_3, X_4) &= (X_1, X_2, X_4, X_3), \\
A_{\sigma_4}(X_1, X_2, X_3, X_4) &= (X_1, X_2, X_3, X_4).
\end{align*}
$$

$\triangle$
Hence, the corresponding fundamental network is given by

\[
\begin{align*}
\dot{X}_1 &= f(X_1, X_2, X_2, X_1), \\
\dot{X}_2 &= f(X_1, X_2, X_1, X_2), \\
\dot{X}_3 &= f(X_1, X_2, X_4, X_3), \\
\dot{X}_4 &= f(X_1, X_2, X_3, X_4).
\end{align*}
\]

This fundamental network has been depicted in Figure 5. It is clear that the maps

\[
\begin{align*}
\pi_1 : (x_1, x_2) &\mapsto (X_1, X_2, X_3, X_4) := (x_1, x_2, x_2, x_1), \\
\pi_2 : (x_1, x_2) &\mapsto (X_1, X_2, X_3, X_4) := (x_1, x_2, x_1, x_2)
\end{align*}
\]

conjugate \( \gamma_f \) to \( \Gamma_f \).

The advantage of studying \( \Gamma_f \) instead of \( \gamma_f \) is that the definition \( \Gamma_f := f \circ A_{\sigma_j} \) explicitly displays the representation of the semigroup \( \Sigma \), whereas the definition \( \gamma_f := f \circ \pi_j \) clearly does not. This has as a consequence that the transformation formulas for the composition and the Lie bracket become completely natural.

**Lemma 10.4** It holds that

\[
\Gamma_f \circ \Gamma_g = \Gamma_{f \circ_{\Sigma} g} \quad \text{and} \quad [\Gamma_f, \Gamma_g] = \Gamma_{[f, g]_{\Sigma}}.
\]

**Proof**: First of all,

\[
\begin{align*}
A_{\sigma_j}(g(A_{\sigma_1}X), \ldots, g(A_{\sigma_n}X)) &= (g(A_{\sigma_{1\sigma_j}}X), \ldots, g(A_{\sigma_{n\sigma_j}}X)) \\
&= (g(A_{\sigma_{1\circ_j}}X), \ldots, g(A_{\sigma_{n\circ_j}}X)).
\end{align*}
\]

This gives that

\[
\begin{align*}
(\Gamma_f \circ \Gamma_g)(X) &= f(\Gamma_f(g(A_{\sigma_1}X), \ldots, g(A_{\sigma_n}X)) = f(g(A_{\sigma_{1\circ_j}}X), \ldots, g(A_{\sigma_{n\circ_j}}X)) = \\
&= f(g(A_{\sigma_1}(A_{\sigma_j}X), \ldots, g(A_{\sigma_n}(A_{\sigma_j}X))) = (f \circ_{\Sigma} g)(A_{\sigma_j}X) = (\Gamma_{f \circ_{\Sigma} g})(X).
\end{align*}
\]

The computation for the Lie bracket is similar.

We stress that Lemma 10.4 holds due to the definition \( \Gamma_f := f \circ A_{\sigma_j} \) and the fact that \( \sigma_j \mapsto A_{\sigma_j} \) is a homomorphism. Lemma 10.4 implies for example that the symbolic
computation of the normal form of $\Gamma_f$ is the same as the symbolic computation of the normal form of $\gamma_f$.

We propose to call $\Gamma_f$ the fundamental network of $\gamma_f$. Two properties make this fundamental network fundamental: first of all, the network architecture of the fundamental network only depends on the multiplicative structure of the semigroup $\Sigma$ and not on the explicit realization of $\Sigma$ itself - in particular, it does not depend on $N$. This means that two semigroup networks have isomorphic fundamental networks if and only if their semigroups are isomorphic. The second fundamental property of the fundamental network is that it is equal to its own fundamental network, if the latter is defined. This follows from Proposition 10.5 below, in which we call a homomorphism of semigroups faithful if it is injective.

**Proposition 10.5** Assume that the homomorphism $\sigma_j \mapsto \tilde{\sigma}_j$ is faithful. Then

$$\tilde{\sigma}_j = \tilde{\sigma}_j$$ and therefore $A_{\sigma_j} = A_{\tilde{\sigma}_j}$ for all $1 \leq j \leq n$.

**Proof:** Recall that $\tilde{\Sigma} = \{\tilde{\sigma}_1, \ldots, \tilde{\sigma}_n\}$ is closed under composition. Thus, the condition that the homomorphism $\sigma_j \mapsto \tilde{\sigma}_j$ from $\Sigma$ to $\tilde{\Sigma}$ is faithful just means that $\tilde{\Sigma}$ is a semigroup. In particular, each map $\tilde{\sigma}_j : \{1, \ldots, n\} \to \{1, \ldots, n\}$ is then well-defined. Now we compute

$$\tilde{\sigma}_j(k) = \tilde{\sigma}_j \circ \tilde{\sigma}_k = \tilde{\sigma}_{\sigma_j(k)} = \tilde{\sigma}_{\sigma_j \circ \sigma_k} = \tilde{\sigma}_{\sigma_j}(k)\tilde{\sigma}_j(k).$$

This proves that $\tilde{\sigma}_j = \tilde{\sigma}_j$ for all $1 \leq j \leq n$ and hence that $A_{\sigma_j} = A_{\tilde{\sigma}_j}$ for all $1 \leq j \leq n$. \qed

Proposition 10.5 brings up the question when the homomorphism $\sigma_j \mapsto \tilde{\sigma}_j$ is faithful, i.e. under which conditions the elements of $\Sigma$ all have different left-multiplicative behavior. We give a partial answer to this question in Remark 10.9 below. The upshot of this remark is that one may essentially always assume the homomorphism to be faithful.

We finish this section with a few simple observations on synchrony and symmetry for $\Gamma_f$. First of all, a direct consequence of Theorem 10.1 is that each $\pi_i \subset V^n$ is an invariant subspace for the dynamics of $\Gamma_f$. Interestingly, another way to see this is by the following

**Proposition 10.6** Every $\pi_i \subset V^n$ is a robust synchrony space for the $\Gamma_f$'s.

**Proof:** Let us define a partition $P$ of $\{1, \ldots, n\}$ by letting $1 \leq j_1, j_2 \leq n$ be in the same element of $P$ if and only if $\sigma_{j_1}(i) = \sigma_{j_2}(i)$. Then

$$\text{Syn}_P = \{X \in V^n | X_{j_1} = X_{j_2} \text{ when } \sigma_{j_1}(i) = \sigma_{j_2}(i)\} = \{(x_{\sigma_1(i)}, \ldots, x_{\sigma_n(i)}) | x \in V^n\} = \text{im } \pi_i.$$

It remains to show that the partition $P$ is balanced for $\tilde{\Sigma}$. This is easy though: when $1 \leq j_1, j_2 \leq n$ are in the same element of $P$, then it holds for all $1 \leq k \leq n$ that

$$\sigma_{\tilde{\sigma}_k(j_1)}(i) = (\sigma_k \circ \sigma_{j_1})(i) = (\sigma_k \circ \sigma_{j_2})(i) = \sigma_{\tilde{\sigma}_k(j_2)}(i),$$

where the middle equality holds because $\sigma_{j_1}(i) = \sigma_{j_2}(i)$. This proves that also $\tilde{\sigma}_k(j_1)$ and $\tilde{\sigma}_k(j_2)$ are in the same element of $P$ and hence that the elements of $\tilde{\Sigma}$ preserve $P$. \qed

Recall that $A_{\sigma_j} \circ \pi_i = \pi_{\sigma_j(i)}$. This implies that $A_{\sigma_j}$ sends the $\Gamma_f$-invariant subspace $\pi_i$ to the $\Gamma_f$-invariant subspace $\pi_{\sigma_j(i)}$. But much more is true: the following result shows that $A_{\sigma_j}$ sends all orbits of $\Gamma_f$ to orbits of $\Gamma_f$, even though $A_{\sigma_j}$ may not be invertible.

**Proposition 10.7**

$$\Gamma_f \circ A_{\sigma_j} = A_{\sigma_j} \circ \Gamma_f.$$

**Proof:**

$$\Gamma_{f \circ A_{\sigma_j}}(X) = f(A_{\sigma_k} \circ A_{\sigma_j}X) = f(A_{\sigma_k \circ \sigma_j}X) = f(A_{\sigma_{\tilde{\sigma}_k(j)}}X) = (\Gamma_f)_{\tilde{\sigma}_k(j)}(X) = (A_{\sigma_j} \circ \Gamma_f)_k(X).$$

The final result of this section shows that $\Gamma_f$ may even have more symmetry: the dynamical input symmetries of $\gamma_f$ are true symmetries of $\Gamma_f$. 23
Proposition 10.8 If\( p \) is a permutation of \( \{1, \ldots, N\} \) and \( q \) is a permutation of \( \{1, \ldots, n\} \) so that \( p \circ \sigma_j = \sigma_{q(j)} \circ p \) for all \( 1 \leq j \leq n \) and if \( f \circ \lambda_q \circ \pi_i = f \circ \pi_i \) for all \( 1 \leq i \leq N \), then
\[
\Gamma_f \circ \lambda_q = \lambda_q \circ \Gamma_f \text{ on every } \operatorname{im} \pi_i.
\]

Proof: Recall that under the conditions of the proposition, it holds that \( \pi_i \circ \lambda_p = \lambda_q \circ \pi_{p(i)} \) and that from this it followed that \( \lambda_q \circ \gamma_f = \gamma_f \circ \lambda_p \). As a consequence,
\[
\Gamma_f \circ \lambda_q \circ \pi_{p(i)} = \Gamma_f \circ \pi_i \circ \lambda_p = \pi_i \circ \gamma_f \circ \lambda_p = \pi_i \circ \lambda_p \circ \gamma_f = \lambda_q \circ \pi_{p(i)} \circ \gamma_f = \lambda_q \circ \Gamma_f \circ \pi_{p(i)}.
\]
Because \( p \) is a permutation, this means that \( \Gamma_f \circ \lambda_q = \lambda_q \circ \Gamma_f \text{ on every } \operatorname{im} \pi_i \).

Remark 10.9 To explain when the homomorphism \( \sigma_j \mapsto \bar{\sigma}_j \) is faithful, we can make the following definition: we say that \( 1 \leq i \leq N \) is a slave for the network \( \Sigma \) if there are no \( 1 \leq j \leq n \) and \( 1 \leq k \leq N \) so that \( \sigma_j(k) = i \). Thus, a slave is a cell that does not act as input for any other cell, not even for itself. The point of this definition is the following:

Proposition 10.10 If \( \Sigma \) has no slaves, then \( \sigma_j \mapsto \bar{\sigma}_j \) is a faithful homomorphism.

Proof: The relation \( \bar{\sigma}_{j_1} = \bar{\sigma}_{j_2} \) means that \( \sigma_{j_1} \circ \sigma_k = \sigma_{j_2} \circ \sigma_k \) for all \( k \). This implies in particular that \( \sigma_{j_1} = \sigma_{j_2} \) on \( \operatorname{im} \sigma_k \) for all \( k \). But if \( \Sigma \) is free of slaves, then \( \bigcup_{j=1}^n \operatorname{im} \sigma_j = \{1, \ldots, N\} \). Hence, \( \sigma_{j_1} = \sigma_{j_2} \).

If a network has slaves, then we can reduce it until no slaves remain. This works as follows: first of all, we remove any slave from the network. Because slaves do not affect the dynamics of other cells, this can be done without any effect on the network dynamics. Removing slaves may create new slaves: these are the cells that acted as inputs only for the original slaves. These new slaves can also be removed, etc. until a network free of slaves remains.

The remaining network may not be defined unambiguously, because some of the maps in \( \Sigma \) may coincide after the removal of the slaves. This happens when distinct maps in \( \Sigma \) differ only at slaves. Such maps can be identified though, while \( f \) must be redefined. In this way, we produce an unambiguous network that is free of slaves. For such a network \( \gamma_f \), the corresponding \( \Gamma_f \) is a true fundamental network.

11 Some examples and their normal forms

In this section we illustrate the methods and results of this paper by computing the normal forms of two coupled cell networks. Keeping things simple, we restrict our attention to synchrony breaking steady state bifurcations in one-parameter families of networks with one-dimensional cells.

11.1 A skew product network

In the first example, we consider the homogeneous skew product differential equations

\[
\begin{align*}
\dot{x}_1 &= f(x_1, x_2; \lambda), \\
\dot{x}_2 &= f(x_2, x_1; \lambda).
\end{align*}
\]

(11.26)

Here \( x_1, x_2 \in \mathbb{R} \) and \( f : \mathbb{R}^2 \times \mathbb{R} \to \mathbb{R} \). As usual, we will denote the right hand side of (11.26) by \( \gamma_f(x_1, x_2; \lambda) \) and we will henceforth assume that

\[
\gamma_f(0, 0; 0) = 0 \text{ and } D_{x} \gamma_f(0, 0; 0) \text{ is not invertible.}
\]

This means that at the parameter value \( \lambda = 0 \), the origin \((x_1, x_2) = (0, 0)\) is a fully synchronous equilibrium point of (11.26) that undergoes a steady state bifurcation. We wish to study the generic nature of this bifurcation. So let us write

\[
f_{0,0}(X_1, X_2) = D_X f(0, 0; 0)(X_1, X_2) = a_1 X_1 + a_2 X_2 \text{ with } a_1, a_2 \in \mathbb{R}.
\]

With this notation, we have that

\[
\text{mat } D_{x} \gamma_f(0, 0; 0) = \begin{pmatrix}
a_1 + a_2 & 0 \\
a_2 & a_1
\end{pmatrix}.
\]

24
We remark that this linearization matrix is semisimple. And moreover that a steady state bifurcation occurs when one of its eigenvalues $a_1 + a_2$ or $a_1$ vanishes.

The obvious but important remark is now that equations (11.26) define a semigroup coupled cell network. The corresponding semigroup consists of $\sigma_1$ and $\sigma_2$, where

$$\sigma_1(1) = 1, \sigma_2(2) = 2 \text{ and } \sigma_2(1) = 1, \sigma_2(2) = 1.$$  

We depicted this network in Figure 6.

Figure 6: A homogeneous skew product network.

The composition table of \{\sigma_1, \sigma_2\} reads

<table>
<thead>
<tr>
<th></th>
<th>$\sigma_1$</th>
<th>$\sigma_2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\circ$</td>
<td>$\sigma_1$</td>
<td>$\sigma_2$</td>
</tr>
<tr>
<td>$\sigma_1$</td>
<td>$\sigma_1$</td>
<td>$\sigma_2$</td>
</tr>
<tr>
<td>$\sigma_2$</td>
<td>$\sigma_2$</td>
<td>$\sigma_2$</td>
</tr>
</tbody>
</table>

From this table, we can read off that

$$A_{\sigma_1}(X_1, X_2) = (X_1, X_2), \quad A_{\sigma_2}(X_1, X_2) = (X_2, X_2),$$

and hence that the symbolic Lie bracket of this network is given by

$$[f, g]_{\Sigma}(X_1, X_2) = D_1 f(X_1, X_2) g(X_1, X_2) + D_2 f(X_1, X_2) g(X_2, X_2) - D_1 g(X_1, X_2) f(X_1, X_2) - D_2 g(X_1, X_2) f(X_2, X_2).$$

Since $D_x\gamma_f(0, 0; 0)$ is semisimple, so is $\text{ad}^{\Sigma}_{f_{0, 0}} : P^{k,l} \rightarrow P^{k,l}$ for every $k \geq -1$ and $l \geq 0$. Its kernel determines the normal form of $f$. It only requires a little computation to check that

$$\text{ad}^{\Sigma}_{f_{0, 0}} : \begin{cases} (X_1 - X_2)^{\alpha} X_2^{\beta} \rightarrow (1 - \alpha) a_1 - \beta(a_1 + a_2)[(X_1 - X_2)^{\alpha} X_2^{\beta}] & \text{for } \alpha \geq 1 \text{ and } \beta \geq 0, \\ (X_2^{\beta} X_1) \rightarrow (1 - \beta)(a_1 + a_2) X_2^{\beta} & \text{for } \beta \geq 0. \end{cases}$$

This formula nicely confirms that $\text{ad}^{\Sigma}_{f_{0, 0}}$ is semisimple. We now consider the two different codimension one cases:

1. When $a_1 + a_2 = 0$ but $a_1 \neq 0$ then the kernel of $D_x\gamma_f(0, 0; 0)$ is tangent to the synchrony space $\{x_1 = x_2\}$. In this case, the kernel of $\text{ad}^{\Sigma}_{f_{0, 0}}$ is spanned by elements of the form $(X_1 - X_2) X_2^\beta$ and $X_2^\beta$, where $\beta \geq 0$. Thus, the general normal form of $f$ is

$$f(X_1, X_2; \lambda) = (X_1 - X_2) F(X_2; \lambda) + G(X_2; \lambda),$$

with $F(X_2; \lambda) = A(\lambda) + O(X_2)$, $G(X_2; \lambda) = B(\lambda) + C(\lambda) X_2 + D(\lambda) X_2^2 + O(X_2^3)$ and $A(0) = a_1, B(0) = C(0) = 0$. The normal form equations of motion become

$$\dot{x}_1 = G(x_1; \lambda), \quad \dot{x}_2 = G(x_1; \lambda) + (x_2 - x_1) F(x_1; \lambda).$$

This implies first of all that the stationary points of the normal form satisfy $x_1 = x_2$ and secondly that $x_1$ solves the equation $G(x_1; \lambda) = B(\lambda) + C(\lambda) x_1 + D(\lambda) x_1^2 + O(x_1^3) = 0$. Under the generic conditions that $B'(0), D(0) \neq 0$, we thus find the saddle node branches

$$x_1 = x_2 = \pm \sqrt{(-B'(0)/D(0)) \lambda + O(\lambda)}.$$
of synchronous steady states. A straightforward stability analysis reveals that one of these branches consists of equilibria that are linearly stable in the direction of the synchrony space, while the other branch consists of unstable points. We remark that the saddle node bifurcation is also generic in codimension one in the context of vector fields without any special structure.

2. When \(a_1 + a_2 \neq 0\) and \(a_1 = 0\), then the kernel of \(\text{ad}_{f_{0,0}}^\Sigma\) is spanned by elements of the form \((X_1 - X_2)^\alpha\), where \(\alpha \geq 1\), and the element \(X_2\). Hence the general normal form of \(f\) is given by

\[
\mathcal{J}(X_1, X_2; \lambda) = (X_1 - X_2)F(X_1 - X_2; \lambda) + A(\lambda)X_2,
\]

with \(F(X_1 - X_2; \lambda) = B(\lambda) + C(\lambda)(X_1 - X_2) + O(X_1 - X_2)^2\) and \(A(0) = a_2, B(0) = 0\).

The normal form differential equations are

\[
\begin{align*}
\dot{x}_1 &= A(\lambda)x_1, \\
\dot{x}_2 &= A(\lambda)x_1 + (x_2 - x_1)F(x_2 - x_1; \lambda).
\end{align*}
\]

This implies that the stationary points of the normal form satisfy \(x_1 = 0\), while either \(x_2 = 0\) or \(x_2\) solves the equation \(F(x_2; \lambda) = B(\lambda) + C(\lambda)x_2 + O(x_2^2) = 0\). Under the generic conditions that \(B'(0), C(0) \neq 0\), we thus find the two steady state branches

\[
x_1 = x_2 = 0 \text{ and } x_1 = 0, x_2 = (-B'(0)/C(0))\lambda + O(\lambda^2).
\]

These branches exchange stability when they cross. This means that the normal form displays a synchrony breaking transcritical bifurcation. Such a bifurcation is not generic in codimension one in the context of vector fields without any special structure, and is hence forced by the network structure. More precisely, it follows from the presence of the invariant synchrony space.

### 11.2 A nilpotent feed-forward network

Next, we consider differential equations with the network structure defined in Example 2.4:

\[
\begin{align*}
\dot{x}_1 &= f(x_1, x_1, x_1; \lambda), \\
\dot{x}_2 &= f(x_2, x_1, x_1; \lambda), \\
\dot{x}_3 &= f(x_3, x_2, x_1; \lambda).
\end{align*}
\]

Here \(x_1, x_2, x_3 \in \mathbb{R}\) and \(f : \mathbb{R}^3 \times \mathbb{R} \to \mathbb{R}\). Again, let us write

\[
f_{0,0}(X_1, X_2, X_3) = D_x f(0, 0, 0, 0)(X_1, X_2, X_3) = a_1X_1 + a_2X_2 + a_3X_3\text{ for } a_1, a_2, a_3 \in \mathbb{R}.
\]

Then it holds that

\[
\text{mat } D_x f(0, 0, 0; 0) = \begin{pmatrix}
0 & 0 & 0 \\
0 & a_2 + a_3 & a_1 \\
0 & a_3 & a_2 + a_3
\end{pmatrix}.
\]

This shows that a steady state bifurcation takes place when either \(a_1 + a_2 + a_3 = 0\) or \(a_1 = 0\).

Moreover, the linearization matrix is not semisimple. In fact, its SN-decomposition reads

\[
\begin{pmatrix}
a_1 + a_2 + a_3 & 0 & 0 \\
a_2 + a_3 & a_1 & 0 \\
a_3 & a_2 & a_1
\end{pmatrix} = \begin{pmatrix}
a_1 + a_2 + a_3 & 0 & 0 \\
a_2 + a_3 & a_1 & 0 \\
a_3 & a_2 & a_1
\end{pmatrix} + \begin{pmatrix}
0 & 0 & 0 \\
0 & 0 & 0 \\
-a_2 & a_2 & 0
\end{pmatrix}.
\]

As a consequence, we should accordingly decompose \(f_{0,0}\) as

\[
f_{0,0} = f_{0,0}^S + f_{0,0}^N\text{ where } f_{0,0}^S(X_1, X_2, X_3) = a_1X_1 + (a_2 + a_3)X_3, f_{0,0}^N(X_1, X_2, X_3) = a_2(X_2 - X_3).
\]

Recalling that for this network the expression for the symbolic bracket is given in Example 5.3, it again requires a little computation to find that

\[
\text{ad}_{f_{0,0}^S}^\Sigma : \begin{cases}
X_1^\gamma \\
(X_1 - X_3)^\alpha (X_2 - X_3)^\beta X_3^\gamma
\end{cases} \mapsto \begin{cases}
(1 - \gamma)(a_1 + a_2 + a_3)X_3^\gamma \\
[(1 - \alpha - \beta)a_1 - \gamma(a_1 + a_2 + a_3)](X_1 - X_3)^\alpha (X_2 - X_3)^\beta X_3^\gamma
\end{cases}\text{ for } \gamma \geq 0,
\]

\[
\text{ad}_{f_{0,0}^N}^\Sigma : \begin{cases}
X_1^\gamma \\
(X_1 - X_3)^\alpha (X_2 - X_3)^\beta X_3^\gamma
\end{cases} \mapsto \begin{cases}
(1 - \gamma)(a_1 + a_2 + a_3)X_3^\gamma \\
[(1 - \alpha - \beta)a_1 - \gamma(a_1 + a_2 + a_3)](X_1 - X_3)^\alpha (X_2 - X_3)^\beta X_3^\gamma
\end{cases}\text{ for } \alpha + \beta \geq 1, \gamma \geq 0.
\]
and similarly that
\[
\text{ad}^E_{f_{0,0}} : \begin{cases}
(X_1 - X_3) \alpha (X_2 - X_3) \beta X_3^\gamma \mapsto \\
-\alpha_2 (X_1 - X_3) \alpha - 1 (X_2 - X_3) \beta + 1 X_3^\gamma \\
(X_1 - X_3) \alpha X_3^\gamma \mapsto \\
a (X_2 - X_3) \alpha X_3^\gamma - \alpha_2 (X_1 - X_3) \alpha - 1 (X_2 - X_3) X_3^\gamma \\
(X_2 - X_3) \beta X_3^\gamma \mapsto 0
\end{cases}
\]
for \(\alpha, \beta \geq 1, \gamma \geq 0\),
\(\alpha \geq 1, \gamma \geq 0\), and \(\beta, \gamma \geq 0\).

Once more, we now consider the two codimension one cases:

1. If \(a_1 + a_2 + a_3 = 0\) and \(a_1 \neq 0\), then the kernel of \(\text{ad}^E_{f_{0,0}}\) is spanned by terms

\((X_1 - X_3)X_3^\gamma, (X_2 - X_3)X_3^\gamma\) and \(X_3^\gamma\) with \(\gamma \geq 0\).

One checks that \(\text{ad}^E_{f_{0,0}}\) vanishes on this kernel, so the general normal form of \(f\) is
\[
\mathcal{T}(X_1, X_2, X_3; \lambda) = (X_1 - X_3)F(X_3; \lambda) + (X_2 - X_3)G(X_3; \lambda) + H(X_3; \lambda),
\]
where \(F(X_3; \lambda) = A(\lambda) + \mathcal{O}(X_3), \ G(X_3; \lambda) = B(\lambda) + \mathcal{O}(X_3), \ H(X_3; \lambda) = C(\lambda) + D(\lambda)X_3 + E(\lambda)X_3^2 + \mathcal{O}(X_3^3)\) and \(A(0) = a_1, B(0) = a_2, C(0) = D(0) = 0\). The normal form equations of motion are
\[
\begin{align*}
\dot{x}_1 &= H(x_1; \lambda), \\
\dot{x}_2 &= (x_2 - x_1)F(x_1; \lambda) + H(x_1; \lambda), \\
\dot{x}_3 &= (x_3 - x_1)F(x_1; \lambda) + (x_2 - x_1)G(x_1; \lambda) + H(x_1; \lambda).
\end{align*}
\]
(11.28)

It follows that the steady states of the normal form satisfy \(x_1 = x_2 = x_3\), where \(x_1\) satisfies \(H(x_1; \lambda) = C(\lambda) + D(\lambda)x_1 + E(\lambda)x_1^2 + \mathcal{O}(x_1^3) = 0\). Under the generic conditions that \(C'(0), D(0) \neq 0\), this yields the fully synchronous saddle node branches
\[
x_1 = x_2 = x_3 = \pm \sqrt{-C'(0)/D(0)} \lambda + \mathcal{O}(\lambda).
\]

Again, one of these branches is stable and the other one is unstable in the direction of the maximal synchrony space.

2. When \(a_1 = 0, a_2 \neq 0\) and \(a_1 + a_2 + a_3 \neq 0\), then \(\ker \text{ad}^E_{f_{0,0}}\) is spanned by the elements
\(X_3\) and \((X_1 - X_3) ^\alpha (X_2 - X_3) ^\beta\) with \(\alpha + \beta \geq 1\).

This time the action of \(\text{ad}^E_{f_{0,0}}\) on \(\ker \text{ad}^E_{f_{0,0}}\) is nontrivial. The only terms in the kernel that are not in \(\im \text{ad}^E_{f_{0,0}}\) are actually those of the form
\[
(X_1 - X_3)^\alpha, X_2 - X_3 \quad \text{and} \quad X_3, \quad \text{with} \quad \alpha \geq 1.
\]

This means that the general normal form of \(f\) is
\[
\mathcal{T}(X_1, X_2, X_3; \lambda) = (X_1 - X_3)F(X_1 - X_3) + A(\lambda)(X_2 - X_3) + B(\lambda)X_3,
\]
where \(F(X_1 - X_3) = C(\lambda) + D(\lambda)(X_1 - X_3) + \mathcal{O}(X_1 - X_3)^2\) and \(A(0) = a_2, B(0) = a_1 + a_2 + a_3, C(0) = 0\). This gives the equations of motion
\[
\begin{align*}
\dot{x}_1 &= B(\lambda)x_1, \\
\dot{x}_2 &= B(\lambda)x_1 + (x_2 - x_1)F(x_2 - x_1; \lambda), \\
\dot{x}_3 &= B(\lambda)x_1 + A(\lambda)(x_2 - x_1) + (x_3 - x_1)F(x_3 - x_1; \lambda).
\end{align*}
\]
(11.29)

Under the generic assumption that \(C'(0), D(0) \neq 0\), we now find three branches of steady states:
\[
x_1 = x_2 = x_3 = 0,
\]
\[
x_1 = x_2 = 0, x_3 = -(C'(0)/D(0)) \lambda + \mathcal{O}(\lambda^2),
\]
(11.30)
\[
x_1 = 0, x_2 = -(C'(0)/D(0)) \lambda + \mathcal{O}(\lambda^2), x_3 = \pm \sqrt{(a_2C'(0)/D(0)^2)} \lambda + \mathcal{O}(\lambda).
\]

This means that our normal form equations undergo a very particular synchrony breaking steady state bifurcation that comprises a fully synchronous trivial branch, a partially synchronous transcritical branch and fully nonsynchronous saddle-node branches. The solutions on these branches exchange stability in a specific way, as for example depicted in Figure 7.
Figure 7: Bifurcation diagram of a codimension-one steady state bifurcation in the normal form of a three cell feedforward network. Pluses and minuses refer to positive and negative eigenvalues in the eigendirections other than the maximal synchrony space. This figure depicts the solutions of formula (11.30) in case $a_2, C'(0), D(0) > 0$.

12 Colored coupled cell networks

In this final section, we describe how our results on homogeneous coupled cell networks generalize to certain non-homogeneous coupled cell networks. So let us imagine a coupled cell network with cells of different types. We will refer to the different types of cells as colors.

More precisely, let us assume that there are $1 \leq C < \infty$ colors and that for every color $1 \leq c \leq C$ there are precisely $N_c$ cells of color $c$. We label the cells of color $c$ by $1 \leq i \leq N_c$ and assume that the state of the $i$-th cell of color $c$ is described by $x_{i}^{(c)} \in V_c$, where $V_c$ is a linear space that depends on $c$.

We furthermore assume that the discrete- or continuous-time evolution of $x_{i}^{(c)}$ is determined by precisely $n_{(1,c)}$ cells of color 1, by $n_{(2,c)}$ cells of color 2, etc. This assumption is made precise in Definition 12.1 below that, although lengthy, is a straightforward generalization of Definition 2.1.

Definition 12.1 For every $1 \leq c, d \leq C$ and every $1 \leq j \leq n_{(d,c)}$, assume there is a map

$$\sigma_j^{(d,c)} : \{1, \ldots, N_c\} \to \{1, \ldots, N_d\}.$$

We denote the collection of these maps by

$$\Sigma := \{\sigma_1^{(1,1)}, \ldots, \sigma_{n_{(1,1)}}^{(1,1)}; \ldots; \sigma_1^{(C,C)}, \ldots, \sigma_{n_{(C,C)}}^{(C,C)}\}.$$

Next, we define for all $1 \leq c \leq C$ and $1 \leq i \leq N_c$ the maps

$$\pi_i^{(c)} : V_1^{N_1} \times \ldots \times V_C^{N_C} \to V_1^{n_{1,c}} \times \ldots \times V_C^{n_{C,c}}$$

by

$$\pi_i^{(c)}(x^{(1)}; \ldots; x^{(C)}) := \left( x^{(1)}_{\sigma_1^{(1,c)}(i)}, \ldots, x^{(1)}_{\sigma_{n_{(1,c)}}^{(1,c)}(i)}; \ldots; x^{(C)}_{\sigma_1^{(C,c)}(i)}, \ldots, x^{(C)}_{\sigma_{n_{(C,c)}}^{(C,c)}(i)} \right).$$

Now assume that $f = (f^{(1)}, \ldots, f^{(C)})$ is a collection of functions, with

$$f^{(c)} : V_1^{n_{1,c}} \times \ldots \times V_C^{n_{C,c}} \to V_c.$$

Then we define $\gamma_f : V_1^{N_1} \times \ldots \times V_C^{N_C} \to V_1^{N_1} \times \ldots \times V_C^{N_C}$ by

$$\gamma_f^{(c)} := f^{(c)} \circ \pi_i^{(c)}$$

for all $1 \leq c \leq C$ and $1 \leq i \leq N_c$.

We say that $\gamma_f$ is a colored coupled cell network map/vector field subject to $\Sigma$. △
It is important to note that only the compositions
\[ \sigma_{j_1}^{c,d} \circ \sigma_{j_2}^{d,c} : \{1, \ldots, N_c\} \to \{1, \ldots, N_e\} \]
are sensibly defined. This inspires the following definition:

**Definition 12.2** We say that \( \Sigma \) is a *semigroupoid* if for every \( 1 \leq c, d \leq C \) and every \( 1 \leq j_1 \leq n_{d,c} \) and \( 1 \leq j_2 \leq n_{c,d} \) there is precisely one \( 1 \leq j_3 \leq n_{c,e} \) such that
\[ \sigma_{j_1}^{c,d} \circ \sigma_{j_2}^{d,c} = \sigma_{j_3}^{e,c} . \]

When a collection \( \Sigma \) as in Definition 12.1 is not semigroupoid, then it generates one: the smallest semigroupoid \( \Sigma' \) containing \( \Sigma \).

**Example 12.3** The completely general \( C \)-dimensional differential equation
\[ x^{(c)} = f^{(c)}(x^{(1)}; \ldots; x^{(C)}) \]
for \( 1 \leq c \leq C \) and \( x^{(c)} \in V_c \)
is an example of a colored coupled cell network with \( C \) colors and one cell of each color. The elements of \( \Sigma = \{ \sigma_1^{(1)}; \ldots; \sigma_C^{(C)} \} \) are all defined by \( \sigma_j^{(d,d)}(1) = 1 \). They obviously form a semigroupoid.

**Example 12.4** The general 2-dimensional skew product differential equation
\begin{align*}
x^{(1)}(1) &= f^{(1)}(x^{(1)}) \\
x^{(2)}(1) &= f^{(2)}(x^{(1)}, x^{(2)})
\end{align*}
with \( x^{(1)} \in V_1 \) and \( x^{(2)} \in V_2 \) is an example of a colored coupled cell network with two colors and one cell of each color. The elements of \( \Sigma = \{ \sigma_1^{(1,1)}; \sigma_1^{(1,2)}; \sigma_1^{(2,1)} \} \) are all defined by \( \sigma_j^{(d,d)}(1) = 1 \) and thus form a semigroupoid. See Figure 8.

![Figure 8: A colored skew product network.](image)

**Example 12.5** The 3-dimensional differential equation
\begin{align*}
x^{(1)}(1) &= f^{(1)}(x^{(1)}_2, x^{(2)}_1) \\
x^{(2)}(1) &= f^{(1)}(x^{(1)}_2, x^{(2)}_1) \\
x^{(3)}(1) &= f^{(2)}(x^{(1)}_2)
\end{align*}
with \( x^{(1)}_1, x^{(1)}_2 \in V_1 \) and \( x^{(2)} \in V_2 \) is an example of a colored coupled cell network with two colors: two cells of color 1 and one cell of color 2. Here, \( \Sigma = \{ \sigma_1^{(1,1)}; \sigma_1^{(1,2)}; \sigma_1^{(2,1)} \} \) where these maps are defined by
\[ \sigma_1^{(1,1)}(1) = 2, \sigma_1^{(1,1)}(2) = 2, \sigma_1^{(1,2)}(1) = 2, \sigma_1^{(2,1)}(1) = 1, \sigma_1^{(2,1)}(2) = 1 . \]
Again, one quickly checks that \( \Sigma \) is a semigroupoid. See Figure 9.

Under the condition that \( \Sigma \) is a semigroupoid, all results of this paper on Lie algebras and normal forms can be generalized to colored coupled cell networks. As an illustration, we state a few facts here without proof.
Figure 9: An example of a colored network with three cells of two colors.

Theorem 12.6 If $\Sigma$ is a semigroupoid, then for each $\sigma_{j}^{(d,c)} \in \Sigma$ there is a unique linear map

$$A_{\sigma_{j}^{(d,c)}} : V_{1}^{n(1,c)} \times \ldots \times V_{C}^{n(C,c)} \rightarrow V_{1}^{n(1,d)} \times \ldots \times V_{C}^{n(C,d)}$$

such that for all $1 \leq i \leq N_{c}$ it holds that

$$A_{\sigma_{j}^{(d,c)}} \circ \pi^{(c)}_{i} = \pi^{(d)}_{\sigma_{j}^{(d,c)}(i)} \cdot$$

These maps satisfy the relations $A_{\sigma_{j_{1}}^{(e,d)}} \circ A_{\sigma_{j_{2}}^{(d,c)}} = A_{\sigma_{j_{1}}^{(e,d)} \circ \sigma_{j_{2}}^{(d,c)}}$ and thus form a representation of the semigroupoid $\Sigma$.

Theorem 12.7 If $\Sigma$ is a semigroupoid, then

$$\gamma_{f} \circ \gamma_{g} = \gamma_{f \circ \Sigma g}$$

in which $(f \circ \Sigma g)^{(c)}$ is equal to

$$f^{(c)} \circ \left( g^{(1)} \circ A_{\sigma_{j_{1}}^{(1,c)}} \times \ldots \times g^{(C)} \circ A_{\sigma_{j_{1}}^{(C,c)}} \right) \cdot$$

Theorem 12.8 If $\Sigma$ is a semigroupoid, then

$$[\gamma_{f}, \gamma_{g}] = \gamma_{[f,g]_{\Sigma}}$$

in which $[f,g]_{\Sigma}^{(c)}$ equals

$$\sum_{d} \sum_{j} \left( D_{\lambda_{j}^{(d)}} f^{(c)} \cdot (g^{(d)} \circ A_{\sigma_{j}^{(d,c)}}) - D_{\lambda_{j}^{(d)}} g^{(c)} \cdot (f^{(d)} \circ A_{\sigma_{j}^{(d,c)}}) \right) \cdot$$

In turn, Theorem 12.7 can be used to prove normal form theorems for colored coupled cell networks. That is, the theorems of Sections 6 and 9 remain true with the word “semigroup” replaced by “semigroupoid”.

We conclude with two results that say that the network symmetries and the robust synchrony spaces of a network remain unchanged by the semigroupoid extension.

Lemma 12.9 Let $\Sigma$ be as in Definition 12.1, not necessarily forming a semigroupoid, and let $p$ be a permutation of the cells so that the restriction $p : \{1, \ldots, N_{c}\} \rightarrow \{1, \ldots, N_{c}\}$ preserves the cells of each color. We say that $p$ is a network symmetry if

$$p \circ \sigma_{j}^{(d,c)} = \sigma_{j}^{(d,c)} \circ p$$

for all $1 \leq c, d \leq C$ and all $1 \leq j \leq n_{(d,c)}$.

This means that $\lambda_{\sigma}$ sends orbits of $\gamma_{f}$ to orbits of $\gamma_{f}$.

Then the collection of network symmetries of $\Sigma$ is the same as the collection of network symmetries of the semigroupoid $\Sigma'$ generated by $\Sigma$. 30
Lemma 12.10 Let $\Sigma$ be as in Definition 12.1, not necessarily forming a semigroupoid, and let $P = \{P^{(1)}, \ldots, P^{(C)}\}$ be a collection of partitions, i.e., for all $1 \leq c \leq C$, we have that $P^{(c)} = \{P^{(c)}_1, \ldots, P^{(c)}_{N_c}\}$ is a partition of $\{1, \ldots, N_c\}$. Then the following are equivalent:

i) The collection of partitions is balanced, i.e., for all $1 \leq c, d \leq C$, all $1 \leq j \leq n_{(d,c)}$ and $1 \leq k_1 \leq r_c$ there exists a $1 \leq k_2 \leq r_d$ so that $\sigma_j^{(d,c)}(P^{(c)}_{k_1}) \subset P^{(d)}_{k_2}$.

ii) The subspace

$$
\text{Syn}_p := \{x \in V^{N_1} \times \ldots \times V^{N_C} | \ x^{(c)}_i = x^{(c)}_{i_2} \text{ when } i_1 \text{ and } i_2 \text{ are in the same element of } P^{(c)}\}
$$

is a robust synchrony space for the networks subject to $\Sigma$.

The collection of robust synchrony spaces of $\Sigma$ is the same as the collection of robust synchrony spaces of the semigroupoid $\Sigma'$ generated by $\Sigma$.

References


