Symmetry- and Input-Cluster Synchronization in Networks

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Abstract

We study cluster synchronization in networks and show that the stability of all possible cluster synchronization patterns depends on a small set of Lyapunov exponents. Our approach can be applied to clusters corresponding to both orbital partitions of the network nodes (symmetry-cluster synchronization) and equitable partitions of the network nodes (input-cluster synchronization.) Our results are verified experimentally in networks of coupled opto-electronic oscillators.
I. INTRODUCTION

Cluster synchronization (CS) in networks of coupled oscillators has been the subject of vast research efforts, see e.g., [1–5]. Recent work [3,4] has elucidated the relation between the symmetries of the network topology and the formation of clusters of synchronized dynamical units in the network. Reference [3] analyzed the formation and stability of synchronized clusters, where these clusters correspond to the orbits of the network symmetry group. Reference [4] extended this study to different cluster synchronization patterns, where all of these are the result of symmetry breakings of the largest possible clusters predicted by the symmetry analysis. Reference [4] also considered the formation of additional cluster synchronization patterns, not predicted by the symmetry analysis, for the case of networks whose connectivity is given in the form of a Laplacian matrix ($L$-networks).

Here we consider the important problem of studying whether any pair of nodes (or subset of nodes) belonging to the same cluster synchronizes or not. Note that this is a more general problem than studying whether an entire cluster synchronizes or not. Comparing with the methods in [5,6], our approach provides essential information that is key to addressing this problem. Moreover, different from the method in [5] our method can be applied to explain the emergence of synchronization for nodes that are not related by symmetries, but receive the same total amounts of inputs from their neighboring nodes in different clusters. In what follows, we thus distinguish between symmetry-cluster synchronization and input-cluster synchronization and relate these two concepts to the network orbital partition and equitable partition, respectively. As we will see, the general rule is that nodes in the same cluster of the equitable partition can synchronize, while nodes in different clusters of the equitable partition cannot synchronize. Our main result is a method that is built on top of the symmetry analysis (orbital partition) and that allows us to determine whether any pair or subset of nodes belonging to a cluster of the equitable partition will synchronize or not.

The paper is organized as follows. In Sec. [II] we introduce the orbital and equitable partitions of the nodes of a network. In Sec. [III] we study the stability of the synchronous solution for nodes belonging to clusters corresponding to the orbital and equitable partitions. In Sec. [IV] we describe the stability of the orbital and equitable clusters and present an experimental verification of their stability using a network of optoelectronic oscillators [7]. Finally in Sec. [V] we discuss the emergence of clusters not predicted by the symmetry analysis.
in the case of networks whose connectivity is given in the form of an adjacency matrix.

II. ORBITAL AND EQUITABLE PARTITIONS OF THE NETWORK NODES

A network can be described by an adjacency matrix \( A = \{A_{ij}\} \), whose entries \( A_{ij} = A_{ji} = 1 \), if node \( j \) is connected to node \( i \) and \( A_{ij} = A_{ji} = 0 \) otherwise. We call a so constructed network an \( A \)-network. We introduce the group \( Aut \) of symmetries of the network topology, where each symmetry permutes the network nodes but leaves the network topology unaltered. From knowledge of all the symmetries, the network can be partitioned into \( M \) orbital clusters \( \mathcal{C}_o \). Mathematically, the orbit of a node \( x \), \( C_o(x) \) is defined as the set of nodes \( y \) such that \( \exists \pi \in Aut : y = \pi \ast x \), where \( \pi \ast x \) denotes the node \( y \) that node \( x \) is mapped to under the action of the symmetry \( \pi \). Under the action of all the network symmetries, the network nodes are partitioned into orbits formed of nodes that permute among one another. For example, the orbits of the network in Fig. 1a are \{1, 2, 3, 4\}, \{5, 6, 7, 8\}, and \{9, 10\}.

Another approach to partition the nodes of a network is based on an equitable partition. In an equitable partition, any two clusters \( C_i^e \) and \( C_j^e \) of the partition are such that a node \( x_i \in C_i^e \) has exactly \( n_{ij} \) neighbors in \( C_j^e \) regardless of the choice of \( x_i \). It is important to note that every orbital partition is also equitable but the converse is not true. Given a network, there may be several equitable partitions; out of these there is one equitable partition that is the coarsest, i.e., with the minimum number of clusters. A scalable numerical algorithm to obtain the coarsest equitable partition of a network is presented in [1]. In what follows we will refer to this as the equitable partition of a network and to the corresponding clusters as the equitable clusters.

The condition for a CS pattern to be flow-invariant is that it corresponds to an equitable network partition, because in that case each node in each cluster receives the same number of inputs from all the clusters.

In Fig. 1a we show an \( A \)-network for which the orbital (Fig. 1a) and the equitable (Fig. 1b) partitions do not coincide. The associated adjacency matrix is shown in Fig. 1c. Note that the green and blue clusters of Fig. 1a merge to form the yellow cluster in Fig. 1b. All nodes of the yellow cluster are connected to two nodes of the red cluster and two other nodes of the yellow cluster. This makes it possible for all the nodes of the yellow cluster...
FIG. 1. Different representations of a network: (a) shows 3 orbital clusters (red, green and blue) for the orbital partition and (b) shows 2 equitable clusters (red and yellow) for the equitable partition. (c) The network adjacency matrix $A$. (d) and (e) show the two quotient networks corresponding to the partitions in (a) and (b), respectively. For each quotient network, the corresponding quotient matrix $Q$ is also shown.

To synchronize. Similarly, all nodes of the red cluster are connected to three nodes of the yellow cluster, allowing all nodes of the red cluster to synchronize.

Given either an equitable or an orbital partition, we define the $M \times M$ quotient matrix $Q$ such that for each pair of clusters $(C^u, C^v)$,

$$Q_{uv} = \sum_{j \in C^v} A_{ij}, \quad i \in C^u, \quad u, v = 1, 2, \ldots M,$$

where $M$ is the number of the clusters of the partition. The network in Fig. 1 has $M = 3$ orbital clusters and $M = 2$ equitable clusters. The two quotient networks are shown in Fig. 1(d-e). As can be seen from Fig. 1(e) each cluster of the full network maps to a single node of the quotient networks. In what follows, when performing the stability analysis, we will
linearize the full network equations about the desired CS solution corresponding to one of these two quotient networks.

III. STABILITY ANALYSIS

A set of general dynamical equations to describe a system of coupled oscillators in discrete time is:

$$\phi_i[n + 1] = F(\phi_i[n]) + \sigma \sum_{j=1}^{N} A_{ij} H(\phi_j[n]),$$  \hspace{1cm} (2) 

\( i = 1, 2, \ldots N \). By grouping together all the nodes in the same clusters, Eqs. (2) become,

$$\tilde{\phi}_u[n + 1] = F(\tilde{\phi}_u[n]) + \sigma \sum_{v}^{M} Q_{uv} H(\tilde{\phi}_v[n]),$$  \hspace{1cm} (3) 

\( u = 1, 2, \ldots M \). Note that Eqs. (3) can be written for both the orbital and equitable network clusters.

We consider here the dynamics of coupled opto-electronic oscillators \[3, 4, 7, 9\] which is modeled by Eqs. (2), where the scalar \( \phi_i[n] \) is the phase of oscillator \( i \) at time \( n \), with \( F(\phi_i) = \beta I(\phi_i), \ H(\phi_i) = I(\phi_i) \) and \( I(\phi_i) = \sin^2(\phi_i + \delta) \). Without loss of generality we set \( \beta = 3.5 \) and \( \delta = \pi/4 \) and leave the coupling parameter \( \sigma \) as a free parameter.
A. Stability of the orbital and equitable quotient network dynamics

An important preliminary step to proceed with our analysis that follows is to characterize stability of the two quotient networks in Fig. 1c and 1d. We linearize the orbital quotient network dynamics (3) about the equitable quotient network state \( \tilde{\phi}_2[n] = \tilde{\phi}_3[n] = \tilde{\phi}_s[n] \). We then study the time evolution of the perturbation \( \tilde{\Delta}\phi = (\tilde{\phi}_2 - \tilde{\phi}_3) \) (for more details see Appendix A), yielding,

\[
\tilde{\Delta}\phi[n+1] = (\beta - \sigma) \sin(2\tilde{\phi}_s[n] + 2\delta) \tilde{\Delta}\phi[n]
\]

In Fig. 2 we show the synchronization error \( |\tilde{\phi}_2 - \tilde{\phi}_3| \) and the maximum Lyapunov exponent (MLE) for Eq. (4) as a function of \( \sigma \). Note that we are then characterizing not only the stability of a partition, but the stability of a partition that behaves in the particular way described by \( \tilde{\phi}_s[n] \). Figs. 4, 5, 7 and 8 show MLE calculations over two different ranges of \( \sigma \): shaded regions (non-shaded regions) correspond to ranges of \( \sigma \) for which the MLE in Fig. 2 is positive and we linearize about the orbital quotient network dynamics (is negative and we linearize about the equitable quotient network dynamics). For the equitable quotient network dynamics, we numerically found that the same attractor is reached independent of the the initial conditions \( \tilde{\phi}_1[0], \tilde{\phi}_2[0] \), when these are randomly chosen in the interval \((0, 2\pi)\). The non shaded regions of Figs. 4, 5, 7, 8 indicate that the MLE’s have been computed around this attractor. For the orbital quotient network dynamics, calculations were performed around the attractor to which Eq. (3) converges when it is evolved from \( \tilde{\phi}_2[0] \approx \tilde{\phi}_3[0] \). The shaded regions of Figs. 4, 5, 7, 8 indicate that the MLE’s have been computed around such an attractor. It is also worth mentioning that it is possible to find a different asymptotic solution for \( \tilde{\phi}_2[n] \neq \tilde{\phi}_3[n] \) with a completely different associated MLE. This however does not affect the results shown in what follows.

Moreover, we have studied the full dynamics of the orbital quotient network and of the equitable quotient network to understand the relation between cluster synchronization and the system’s behavior. Interestingly, we have seen that the MLEs are sensitive to whether the asymptotic quotient network behavior is a fixed point, a limit cycle or a strange attractor. The drops in the MLEs are in fact related with a blue-sky catastrophe, a bifurcation in which the strange attractor on which the system evolves is destroyed and is replaced by a simpler attractor (in our case, a period 2 limit cycle). On the other hand, the smooth increases of
the MLEs are related with a smooth complexification of the asymptotic trajectory due to a Feigenbaum cascade (i.e., a sequence of period doubling bifurcations).

Next we analyze stability of all the valid CS states for Eqs. (2). Using the technique in [3] we compute the transformation matrix $T$ (non-normalized) of the network based on the irreducible representations (IRRs) of the symmetry group. A code that performs these operations and outputs the matrix $T$ is provided in [10]. This transformation matrix $T$ maps from the node coordinate system into the IRR coordinate system [3]. Moreover, $TAT^{-1} = B$, where $B$ is a block diagonal matrix. The trivial representation (rows of the matrix $T$ are such that $T_{ij} = 1$ if node $j$ is in cluster $i$ and $T_{ij} = 0$ otherwise) corresponds to the synchronous state and we place those rows at the top of the matrix $T$, which means the first $M \times M$ block of the matrix $B$ corresponds to perturbations parallel to the synchronization manifold. The remaining transverse blocks correspond to perturbations orthogonal to the synchronization manifold [3]. Following [3], when studying stability of the CS state, a set of decoupled variational equations can be written, each one corresponding to a block of the matrix $B$ (see Appendix B). The transverse blocks are those responsible for stability of the CS solution.

Referring to Fig. 3, it is important to note that each row of the matrix $T$ corresponds to a block of the matrix $B$ and so to a MLE. From these MLEs we can predict stability of every possible CS pattern. For the network in Fig. 1a, the first three rows of the transformation matrix $T$ correspond to the three orbital partition clusters (the trivial representation). The fourth row corresponds to the transverse block $\{-1\}$ of the matrix $B$ and the associated MLE is $\lambda_1$. The fifth row corresponds to the transverse block $\{+1\}$ of the matrix $B$ and the associated MLE is $\lambda_2$. Rows 6 to 8 correspond to the 3 by 3 transverse block of the matrix $B$, and the associated MLE is $\lambda_3$. This block determines stability about the transverse symmetry axis in Fig. 4a of all three clusters, which are therefore intertwined [3]. Rows 9 and 10 correspond to the transverse block $\{0\}$, with degeneracy 2, of the matrix $B$, and the associated MLE is $\lambda_4$.

Figure 4 contains a graphical representation of how the rows of the matrix $T$ correspond to several symmetry breakings. The inner circles inside each node for the graphs on the left side of Fig. 4 describe such symmetry breakings. Two inner circles are colored the same (differently) if the nodes remain synchronized (do not remain synchronized) after the symmetry breaking. The condition for the symmetry breaking to occur is that the corresponding MLE is positive. The plots on the right hand side of Fig. 4 show the MLE
FIG. 3. The $T$ matrix based on the orbital partition and the corresponding $B$ matrix. Coloring is reminiscent of the colors used in Fig. 1a, namely pink rows correspond to the red cluster, light green rows to the green cluster and blue rows to the blue cluster.

Knowledge of these MLEs (i.e., the plots on the rhs of Fig. 4 for our example network) is all we need to compute stability of all the possible CS patterns. For example, we can assess whether synchronization of any pair of nodes in a cluster or synchronization of any cluster is stable or not. However, while the number of CS patterns can be extremely large [4], the
number of these MLEs is upper bounded by $N - M$, the number of transverse directions. In the case in which stability of several CS patterns needs to be checked, this comes with an important computational advantage as the MLE calculations do not need to be repeated. As we will see, the matrix $T$ provides the connection between the MLEs and all the CS patterns.

Next we show that the transformation matrix $T$ contains essential information that can be used to assess stability of the synchronous state for different pairs of nodes (which can be generalized to any subset of nodes). We start from the assumption that nodes in different clusters of the orbital partition cannot synchronize [4] (as we will see this assumption turns out being incorrect for the case in which the orbital partition and the equitable partition do not coincide.) Then we focus on understanding whether any two nodes in a cluster of the orbital partition synchronize or not. The matrix $T$ can be used as a table to check whether any pair of nodes in the same cluster, say $i$ and $j$, synchronize or not based on the following steps,

- Select columns $i$ and $j$ of the matrix $T$.
- Find all the rows $k$ of the matrix $T$ such that $T_{ki} \neq T_{kj}$. For each such row, consider the corresponding MLEs.

The condition for the two nodes $i$ and $j$ to synchronize is that all the Lyapunov exponents of the blocks corresponding to the rows $k$ for which $T_{ki} \neq T_{kj}$ are negative. Or analogously that the maximum of all those Lyapunov exponents is negative. For example, to compute the synchronization error between nodes 5 and 7, the rows of the matrix $T$ with different entries are 5 and 8, hence the MLEs are $\lambda_2$ and $\lambda_3$. Then the condition for nodes 5 and 7 to synchronize is that $\max(\lambda_2, \lambda_3) < 0$. Following the same method, it is easy to see that the condition for the entire green cluster to synchronize is that $\max(\lambda_1, \lambda_2, \lambda_3) < 0$.

Using this approach, we have determined the stability of different pairs of nodes belonging to the same cluster. The stability of the cluster states is verified by implementing the network from Fig. 1a in the experimental system described in detail in ref. [7] as well as by direct simulation of Eq. 2. Figure 5 shows synchronization errors from both experimental measurements and numerical simulations and the associated MLEs for all the possible node pairs that can arise in the orbital partition. Because $\phi$ is defined modulo $2\pi$, we define
FIG. 4. Plots a, c, e and g show different symmetry breakings of the network orbital partition. The inner circles inside each node describe such symmetry breakings. Two inner circles are colored the same (differently) if the nodes remain synchronized (do not remain synchronized) after the symmetry breaking. Plots b, d, f and h show the MLE vs. the coupling strength $\sigma$ corresponding to the different symmetry breakings on the left (a, c, e, g). shaded regions (non-shaded regions) indicate ranges of $\sigma$ for which the MLE in Fig. 2 is positive (negative.)
synchronization error between two nodes $i$ and $j$ as $e \equiv \min(\langle |\phi_i - \phi_j| \rangle, \langle ||\phi_i - \phi_j| - 2\pi|| \rangle)$,
where $\langle \cdot \rangle$ indicates an average over time.

The experimental measurements validate the MLE predictions and the direct numerical simulations. The shaded regions in the figure indicate the ranges of $\sigma$ for which the MLE in Fig. 2 is positive. We also repeated our numerical simulations by adding a small amount of noise to the dynamics, for which we observed improved agreement with the $\sigma$-ranges of stable synchronization between experiment and simulation (not shown).

We now consider the stability problem applied to the equitable partition. In Fig. 2 we observe that for certain ranges of $\sigma$, the equitable quotient network dynamics is stable, which suggests the possibility that pairs of nodes from the green and blue clusters can synchronize. By following the method in [4], we can now construct a new transformation matrix $T'$ that corresponds to the network equitable partition. Namely, we ‘merge’ the green and blue clusters (Fig. 3a), so to generate a new synchronous vector \[0 0 0 0 1 1 1 1 1 1\] and an orthogonal vector \[0 0 0 0 1 1 1 1 2 -2\] in the plane determined by the first and third vectors/rows of the matrix $T$. In Fig. 6 we present the new transformation matrix $T'$ and the corresponding matrix $B' = T'AT'^{-1}$.

As can be seen, when the clusters are merged the dimension of the parallel block decreases by 1 and that of the transverse block increases by 1. A new transverse block is generated with associated MLE $\lambda'_1$. It is interesting to observe that in this case the new MLE obeys the same equation as $\lambda_1$ in Fig. 4d, hence $\lambda'_1 = \lambda_1$.

Since it is possible for nodes in the green and in the blue clusters to synchronize, then it makes sense to study whether pairs of nodes belonging to these clusters synchronize or not. In order to be able to perform this study, we follow the same procedure outlined before, but using the new table $T'$. Namely, for any pair of nodes $i, j$, we need to check whether the maximum of the MLEs corresponding to the rows $k$ of the matrix $T'$ such that $T'_k i \neq T'_k j$ is positive or negative. In Fig. 7 we show the synchronization error for the pairs of nodes (5, 9) and (5, 10) along with the associated MLEs. The experiments provide a validation of the theory.

### IV. SYNCHRONIZATION OF THE ORBITAL AND EQUITABLE CLUSTERS

Here we describe how the symmetry breakings affect each one of the three clusters in Fig. 1a. The red cluster can split into two pairs of nodes $\phi_1 = \phi_2$ and $\phi_3 = \phi_4$, which
FIG. 6. The transformation matrix $T'$ corresponding to the equitable partition and the associated block diagonal matrix $B'$. Coloring is reminiscent of the colors used in Fig. 1b, namely pink rows correspond to the red cluster, light yellow rows to the yellow cluster.

corresponds to a symmetry breaking occurring along the transverse symmetry axis (Fig. 4a). There are two other possible CS patterns of the red cluster: either $\phi_1 = \phi_3$ and $\phi_2 = \phi_4$ or $\phi_1 = \phi_4$ and $\phi_2 = \phi_3$. The double transverse block $\{0\}$ describes stability of these two cluster synchronization patterns of the red cluster, so that depending on the associated MLE, they are either simultaneously stable or simultaneously unstable. The top and bottom bi-directional arrows in Fig. 4g represent the possibility of swapping nodes 3, 4 and 1, 2 in an arbitrary way.

Similarly, the green cluster can split into two pairs of nodes $\phi_5 = \phi_8$ and $\phi_6 = \phi_7$ which correspond to a symmetry breaking along the transverse symmetry axis (Fig. 4a). As can
FIG. 7. Node-pairs synchronization errors (blue is from simulation and green is from experiment) and associated MLEs (in orange). Shaded regions indicates the range of σ for which the MLE in Fig. 2 is positive.

be seen in Fig. 4c and Fig. 4e, the green cluster can also break into two other possible pairs: either φ₅ = φ₇ and φ₆ = φ₈, which depends on the transverse block {−1} or φ₅ = φ₆ and φ₇ = φ₈, which depends on the transverse block {+1}. Finally the blue cluster can break along the transverse symmetry axis in Fig. 4a, which depends on the transverse 3 × 3 block alone.

Similarly to what shown for pairs of nodes, the matrix $T$ ($T'$) can be used as a table to check whether any subset of nodes from the same cluster synchronizes or not based on the following steps,

- Say the subset of nodes from the same cluster of the orbital (equitable) partition is formed of nodes $i_1, i_2, ..., i_p$. Select columns $i_1, i_2, ..., i_p$ of the matrix $T$ ($T'$).

- Find all the rows $k$ of the matrix $T$ ($T'$) such that at least one of the entries $T_{ki_1}, T_{kji_2}, T_{kji_p}$ ($T'_{ki_1}, T'_{kji_2}, T'_{kji_p}$) is different from the others. For such rows, consider the corresponding MLEs.

The condition for the subset of nodes $i_1, i_2, ..., i_p$ to synchronize is that all the MLEs of the blocks corresponding to the rows $k$ for which at least one of the entries $T_{ki_1}, T_{kji_2}, T_{kji_p}$ ($T'_{ki_1}, T'_{kji_2}, T'_{kji_p}$) is different from the others, are negative. Or analogously that the maximum
of all those Lyapunov exponents is negative. Obviously, this approach applies also to the case that the subset of nodes coincides with the whole cluster.

Figure 8 shows the cluster synchronization errors, computed as the averaged pairwise synchronization error over all the possible pairs inside a cluster, and associated MLEs for the red cluster, green cluster, blue cluster, and yellow cluster, and again we see agreement between our stability predictions and the synchronization errors. The shaded regions in the figure indicate the ranges of $\sigma$ for which the MLE in Fig. 2 is positive.

V. EQUITABLE PARTITIONS OF $A$-NETWORKS

Different from what was previously reported [4], here we have shown that the formation of cluster synchronous patterns not predicted by the symmetry analysis is also possible in
the case of networks whose connectivity is given in the form of an adjacency matrix (A-networks).

Based on previous work from some of us [3], the network symmetries can be used to partition the nodes of an A-network into clusters corresponding to the orbits of the symmetry group: nodes in the same cluster can synchronize, while nodes in different clusters cannot synchronize. This statement does not hold in the case that the clusters predicted by the symmetry analysis do not coincide with those predicted based on the inputs each node receives [1, 6]. It is indeed possible that two (or more) nodes may synchronize if they receive the same inputs from their neighboring nodes, even when they are not related by a symmetry relation. When this happens, one should instead consider the equitable partition which corresponds to a refinement of the orbital partition provided by the symmetry analysis [1, 6]. We can then conclude that nodes in the same cluster of the equitable partition can synchronize, while nodes in different clusters of the equitable partition cannot synchronize. Then the question arises whether any two nodes belonging to the same cluster of the equitable partition will synchronize or not. In this paper, we have proposed a method that provides an answer to this question. This method allows us to determine whether any pair of nodes belonging to a cluster of the equitable partition will synchronize or not. We have shown that though the symmetry analysis cannot be directly applied to analyze stability of these patterns, it can be conveniently extended to address this problem.

VI. CONCLUSIONS

In this paper we have proposed a unified approach to determine whether any subset of nodes belonging to a cluster of the equitable partition synchronizes or not. One of the most useful quantities to come out of the computational group theory analysis of a network for orbital and equitable clusters is the transformation matrix $T$. Not only does the matrix $T$ block diagonalize the coupling matrix, but the rows of $T$ are the bases of the IRR of the network symmetry group, so that each row shows the directions of the perturbations associated with that IRR. These are the directions each cluster’s nodes will deviate in a symmetry breaking bifurcation. This is essential knowledge for obtaining normal forms for the bifurcation analysis and for identifying the new clusters created. Note, this holds for equitable clusters and Laplacian clusters [4], too. Our results indicate a tradeoff between compu-
tational complexity [5] and completeness of information in addressing the problem of CS stability in networks, in that lower computational complexity may result in less information about the stability of the symmetry breakings.

**Appendix A: Stability of the two Quotient Networks Dynamics**

In order to study stability of the full network dynamics, we first need to characterize stability of the two quotient networks in Fig. 1d and 1e. We write down equation (3) for the quotient network in Fig. 1d,

\[ \tilde{x}_1[n + 1] = F(\tilde{x}_1[n]) + \sigma (2H(\tilde{x}_2[n]) + H(\tilde{x}_3[n])) \]

\[ \tilde{x}_2[n + 1] = F(\tilde{x}_2[n]) + \sigma (2H(\tilde{x}_1[n]) + H(\tilde{x}_2[n]) + H(\tilde{x}_3[n])) \]

\[ \tilde{x}_3[n + 1] = F(\tilde{x}_3[n]) + \sigma (2H(\tilde{x}_1[n]) + 2H(\tilde{x}_2[n])) \]  

(A1)

We are interested in studying stability of the equitable quotient network dynamics. The reason is that in order to analyze the CS of the full network, we are going to linearize the full network equations about either the quotient network dynamics for the equitable partition or for the orbital partition. Thus we linearize Eq. (A1) about the state \( \tilde{x}_s = \tilde{x}_2 = \tilde{x}_3 \) corresponding to the equitable partition quotient network dynamics, yielding

\[ \delta \tilde{x}_1[n + 1] = D F(\tilde{x}_s) \delta \tilde{x}_1[n] + \sigma D H(\tilde{x}_s) [2\delta \tilde{x}_2[n] + \delta \tilde{x}_3[n]] \]  

\[ \delta \tilde{x}_2[n + 1] = D F(\tilde{x}_s) \delta \tilde{x}_2[n] + 2\sigma D H(\tilde{x}_s) \delta \tilde{x}_1[n] + \sigma D H(\tilde{x}_s) [\delta \tilde{x}_2[n] + \delta \tilde{x}_3[n]] \]

\[ \delta \tilde{x}_3[n + 1] = D F(\tilde{x}_s) \delta \tilde{x}_3[n] + \sigma [2D H(\tilde{x}_s) \delta \tilde{x}_1[n] + 2D H(\tilde{x}_s) \delta \tilde{x}_2[n]] \]  

(A2a)

(A2b)

(A2c)

We define \( \Delta \tilde{x}_s[n] = (\delta \tilde{x}_2[n] - \delta \tilde{x}_3[n]) \) and we write,

\[ \Delta \tilde{x}_2[n + 1] = [D F(\tilde{x}_s) - \sigma D H(\tilde{x}_s)] \Delta \tilde{x}_2[n], \]  

(A3)

which is the same as Eq. (4).

We note that Eq. (A3) does not depend on \( \tilde{x}_1 \), which leads to the general rule for merging two clusters or quotient nodes: both clusters or quotient nodes must receive exactly the same coupling from each other cluster or quotient node in the orbital partition. Note that in the orbital partition quotient network there is only one node for each cluster. Then a condition for two nodes of the orbital quotient network to merge is that they receive exactly the same coupling from each other node of the quotient network.
Based on this analysis, if for a certain choice of parameters the quotient network for
the equitable partition is found to be unstable, one would not linearize about this state.
Conversely, if it is found to be stable, we know that linearizing about the orbital network
partition may actually result in linearizing about the equitable network partition (depending
on the choice of the initial conditions).

Appendix B: Decoupling the variational equations

Consider the orbital partition of the network nodes in \( M \) clusters, \( C_m, m = 1, ..., M \), with
synchronous motion \( s_m[n] \). Then we can write the original vectorial variational equations
about the synchronized solutions in the node coordinates,

\[
\delta x[n + 1] = \left[ \sum_{m=1}^{M} E^{(m)} \otimes DF(s_m[n]) + \sigma A \sum_{m=1}^{M} E^{(m)} \otimes DH(s_m[n]) \right] \delta x[n],
\]

where the \( Nn \)-dimensional vector \( \delta x[n] = [\delta x_1[n]^T, \delta x_2[n]^T, ..., \delta x_N[n]^T]^T \) and \( E^{(m)} \) is an
\( N \)-dimensional diagonal matrix such that

\[
E^{(m)}_{ii} = \begin{cases} 
1, & \text{if } i \in C_m, \\
0, & \text{otherwise},
\end{cases}
\]

where \( i = 1, ..., N \). Note that \( \sum_{m=1}^{M} E^{(m)} = I_N \), where \( I_N \) is the \( N \)-dimensional identity matrix.

Applying \( T \) to Eq. (B1) we arrive at the variational matrix equation shown in Eq. (B3), where
\( \eta(n) = T \otimes I_n \delta x(n) \), \( J^{(m)} \) is the transformed \( E^{(m)} \), and \( B \) is the block diagonalization of
the coupling matrix \( A \),

\[
\eta[n + 1] = \left[ \sum_{m=1}^{M} J^{(m)} \otimes DF(s_m[n]) + \sigma B \otimes I_n \sum_{m=1}^{M} J^{(m)} \otimes DH(s_m[n]) \right] \eta[n],
\]

where we have linearized about synchronized cluster states \( \{s_1, ..., s_M\} \), \( \eta[n] \) is the vector
of variations of all nodes transformed to the IRR coordinates and \( DF \) and \( DH \) are the Jacobians of the nodes’ vector field and coupling function, respectively.

We can write the block diagonal \( B \) as a direct sum \( \bigoplus_{l=1}^{L} I_{d(l)} \otimes B_l \), where \( B_l \) is a \( p_l \times p_l \) matrix with \( p_l \) = the multiplicity of the \( l \)th IRR in the permutation representation, \( L \) = the
number of IRRs present, and \( d(l) \) = the dimension of the \( l \)th IRR, so that \( \sum_{l=1}^{L} d(l)p_l = N \).
The trivial representation \( l = 1 \) which is associated with the motion in the synchronization manifold has \( p_1 = M \). The remaining transverse blocks all decouple in independent variational equations that determine stability of the \( CS \) solution.

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[1] I. Belykh and M. Hasler, Chaos: An Interdisciplinary Journal of Nonlinear Science 21, 016106 (2011).
[2] P. Ji, T. K. D. Peron, P. J. Menck, F. A. Rodrigues, and J. Kurths, Physical review letters 110, 218701 (2013).
[3] L. Pecora, F. Sorrentino, A. Hagerstrom, T. Murphy, and R. Roy, Nature communications 5, 4079 (2014).
[4] F. Sorrentino, L. M. Pecora, A. M. Hagerstrom, T. E. Murphy, and R. Roy, Science advances 2, e1501737 (2016).
[5] Y. S. Cho, T. Nishikawa, and A. E. Motter, Physical review letters 119, 084101 (2017).
[6] M. T. Schaub, N. O’Clery, Y. N. Billeh, J.-C. Delvenne, R. Lambiotte, and M. Barahona, Chaos: An Interdisciplinary Journal of Nonlinear Science 26, 094821 (2016).
[7] J. D. Hart, D. C. Schmadel, T. E. Murphy, and R. Roy, Chaos: An Interdisciplinary Journal of Nonlinear Science 27, 121103 (2017).
[8] A. J. Schwenk, in Graphs and Combinatorics (Springer, 1974) pp. 153?172.
[9] C. R. Williams, T. E. Murphy, R. Roy, F. Sorrentino, T. Dahms, and E. Schöll, Physical review letters 110, 064104 (2013).
[10] A. Hagerstrom, “Network symmetries and synchronization (https://sourceforge.net/projects/networksym/),” (2014).