Mediated Remote Synchronization: the Number of Mediators Matters

Yuzhen Qin, Ming Cao, Brian D.O. Anderson, Danielle S. Bassett, and Fabio Pasqualetti

Abstract—Cortical regions without direct neuronal connections have been observed to exhibit synchronized dynamics. A recent empirical study has further revealed that such regions that share more common neighbors are more likely to behave coherently. To analytically investigate the underlying mechanisms, we consider that a set of \( n \) oscillators, which have no direct connections, are linked through \( m \) intermediate oscillators (called mediators), forming a complete bipartite network structure. Modeling the oscillators by the Kuramoto-Sakaguchi model, we rigorously prove that mediated remote synchronization, i.e., synchronization between those \( n \) oscillators that are not directly connected, becomes more robust as the number of mediators increases. Simulations are also carried out to show that our theoretical findings can be applied to other general and complex networks.

I. INTRODUCTION

Synchronization has been pervasively observed in the human brain and is believed to facilitate neuronal communication between cortical regions [1]. Two synchronized regions of neuronal ensembles are more likely to communicate efficiently since synchronization creates temporal windows for interaction [2]. Different cognitive tasks usually require a different set of cortical regions to communicate, and the communication structure is modulated by dynamical patterns of synchronization [3]. Moreover, transient patterns of synchrony can subserve information routing between cortical regions [4]. The underlying anatomical brain network has been shown to play a fundamental role in shaping various patterns of synchrony [5]. However, the mechanisms whereby these patterns arise from the anatomical substrate remains poorly understood. Empirical findings suggest that directly and tightly connected regions in the anatomical network are more likely to exhibit coherent dynamics [6]. Nevertheless, strong evidence reveals that cortical regions without direct axonal links can also show synchrony [7]. Such synchronization is referred to as remote synchronization. Morphological symmetry in the anatomical network is a mechanism besides some others, e.g., cytoarchitectonic similarity [8] and gene co-expression [9], that are believed to account for the emergence of remote synchronization. Cortical regions located at symmetric positions mirror their functionality [10].

It is shown in [11] that two distant neuronal regions symmetrically connected through a third one surprisingly display zero-lag synchronization even in the presence of large synaptic delays. The third region, acting as a relay or mediator, plays a crucial role. A recent empirical study further shows that the level of synchrony between two remote regions significantly correlates with the number of such relays or mediators in the anatomical network [12]. However, a theoretical explanation is still missing, which motivates us to analytically investigate the effect of the number of mediators on remote synchronization. To this aim, we consider a complete bipartite network of oscillators, which consists of two disjoint sets of sizes \( n \) and \( m \), respectively. We investigate how stable remote synchronization can arise between the set of \( n \) oscillators through the mediation of another set, which we refer to as mediators following [13].

Related work: While complete synchronization has been extensively studied (see [14] for a survey), some attention has recently also been paid to partial or cluster synchronization due to its broad applications [15]–[18]. As a particular form of partial synchronization, remote synchronization has also attracted considerable interest, e.g., [19], [20]. Yet rigorous analytical studies, despite some attempts [21], remain marginal. In contrast to most of the existing analytical works on remote synchronization, we use the Kuramoto-Sakaguchi model [22] to describe the dynamics of the oscillators and study the stability of remote synchronization. Compared to the classic Kuramoto model [23], there is an additional phase shift term in the Kuramoto-Sakaguchi model, which has been used to model synaptic delays [24].

Contributions: The contribution of this paper is fourfold. First, it is the first attempt, to the best of our knowledge, to theoretically study remote synchronization in a complete bipartite network (a simulation-based study has been performed [25]). Second, we show that the stability of remote synchronization depends crucially on the phase shift, for which a threshold is identified. A phase shift beyond this threshold can prevent stable remote synchronization. Moreover, this threshold increases with the number of mediators, indicating that more mediators make remote synchronization more robust against phase shifts. This observation provides a rigorous explanation for the simulated and empirical findings in [12], bringing a deeper understanding of the role that the anatomical network plays in shaping patterns of
synchrony in the brain. Third, in sharp contrast to most of the existing results, e.g., [26], [27], which only provide sufficient conditions for the existence of exponentially stable frequency synchronization, we present an almost sufficient and necessary condition. Fourth and finally, we find through simulations that remote synchronization remains stable for any phase shift if the number of mediators exceeds that of the mediated oscillators. Also, our simulation results show that bipartite components in more complex networks play important roles in facilitating robust remote synchronization.

Paper organization: The remainder of this paper is organized as follows. The considered problems are formulated in Section II. Our main results are provided in Section III. The considered problems are formulated in Section II. Our main results are provided in Section III. Some simulation studies are presented in Section IV. Finally, concluding remarks are offered in Section V.

Notation: Let \( \mathbb{R} \), \( \mathbb{R}^+ \), and \( \mathbb{N} \) denote the sets of reals, positive reals, and positive integers, respectively. Given any \( m \in \mathbb{N} \), let \( \mathcal{N}_m = \{1, 2, \ldots, m\} \), and let \( 1_m, 0_m \) and \( I_m \) denote the \( m \)-dimensional all-one vector, all-zero vector, and identity matrix, respectively. Let the unit circle be denoted by \( S^1 \), a point of which is phase. Let \( S^m \) denote the \( m \)-torus.

II. Problem Formulation

Consider a network of \( N \) coupled oscillators whose dynamics are described by

\[
\dot{\theta}_i = \omega + \sum_{j=1}^{N} a_{ij} \sin(\theta_j - \theta_i - \alpha),
\]

where \( \theta_i \in S^1 \) are the phases of the oscillators; \( \omega \in \mathbb{R} \) is the homogeneous natural frequency; \( a_{ij} \) is the coupling strength between oscillators \( i \) and \( j \); and \( \alpha \in (0, \pi/2) \) is the phase shift. Let the graph \( G = (\mathcal{V}, A) \) describe the network structure, where \( \mathcal{V} = \{1, \ldots, N\} \) is the collection of the nodes, and the weighted adjacency matrix \( A = [a_{ij}] \) describes the edges and their weights (there is an edge of weight \( a_{ij} \) between oscillators \( i \) and \( j \) if \( a_{ij} > 0 \)). In the presence of \( \alpha \), complete synchronization is usually not possible. However, it has been shown that oscillators located at morphologically symmetric positions in the network, despite not being directly connected, can be synchronized. This phenomenon is called remote synchronization [19]. If the phase shift \( \alpha \) is small, then remote synchronization appears to be stable; otherwise, it becomes unstable [10].

In this paper we let \( G \) be a complete bipartite graph (see Fig. 1(a)). The dynamics of the oscillators coupled by the network described by \( G \) then become

\[
\dot{\theta}_i = \omega + \sum_{q=1}^{m} a_{iq} \sin(\theta_q - \theta_i - \alpha), i \in \mathcal{N}_m, \tag{2a}
\]

\[
\dot{\theta}_{rp} = \omega + \sum_{j=1}^{n} a_{rpj} \sin(\theta_j - \theta_{rp} - \alpha), p \in \mathcal{N}_m, \tag{2b}
\]

where\(^1\) \( m < n \) and \( n + m = N \). The peripheral oscillators, \( 1, \ldots, n \), are connected via some intermediate oscillators (colored red in Fig. 1(a)). We call those intermediate oscillators mediators, since they are mediating the dynamics of the peripheral oscillators. The peripheral oscillators are called mediated oscillators. Following [13], we also refer to the synchronization of mediated oscillators, \( 1, 2, \ldots, n \), as mediated remote synchronization. When there is only 1 mediator, the network reduces to a star (see Fig. 1(b)). A threshold of the phase shift \( \alpha \), beyond which mediated remote synchronization becomes unstable, has been obtained in [28] for a star network with two mediated oscillators.

We aim to extend this result to a general case in which there can be more than 2 mediated oscillators (i.e., \( n \geq 2 \)). Interestingly, we also allow for more than 1 mediator (i.e., \( m \geq 1 \)) and study how the number of mediators affects the threshold for stability on the phase shift \( \alpha \).

Let \( \theta = (\theta_1, \ldots, \theta_n, \theta_{r1}, \ldots, \theta_{rm})^T \), and for any \( i \) and \( p \) denote the right-hand sides of (2a) and (2b) by \( f_i(\theta) \) and \( g_p(\theta) \), respectively. Then, (2) can be rewritten as \( \dot{\theta}_i = f_i(\theta), \dot{\theta}_{rp} = g_p(\theta) \). In the rest of this paper, we assume that all the coupling strengths are have unit value for simplicity, i.e., \( a_{irp} = 1 \) for all \( i \) and \( r_p \). Under this assumption, the mediated oscillators are ensured to be located at symmetric positions, since swapping their positions does not change the functioning of the overall system.

Next, let us first define the (mediated) remote synchronization manifold, denoted by \( M \). For \( \theta \in S^N \), define \( M := \{\theta \in S^N : \theta_i = \theta_j, \forall i, j \in \mathcal{N}_m\} \). A solution \( \theta(t) \) to (2) is said to be remotely synchronized if \( \theta(t) \in M \) for all \( t \geq 0 \). Note that the phases \( \theta_i(t) \) are not required to equal \( \theta_{rp}(t) \) for all \( t \geq 0 \) in a remotely synchronized solution. We also say that remote synchronization has taken place if a solution to (2) is remotely synchronized.

Remote synchronization is categorized into two types, depending on whether the phases are locked. A solution is phase-locked if every pair-wise phase difference involving the mediators is constant, or, equivalently, when all the frequencies are synchronized. In contrast, the mediators are allowed to have different frequencies from \( \dot{\theta}_i \)’s in the case of phase-unlocked remote synchronization. We are exclusively interested in studying phase-locked remote synchronization in this paper, and thus we refer to it just as remote synchronization for brevity. The (phase-locked) remote synchronization manifold is then defined as follows.

Definition 1: (Remote Synchronization Manifold) For \( \theta \in S^N \), the remote synchronization manifold is defined by \( M_\theta := \{\theta \in M : f_i(\theta) = g_p(\theta), \forall i \in \mathcal{N}_m, p \in \mathcal{N}_m\} \).

It then suffices to identify the threshold of \( \alpha \) beyond which the manifold \( M_\theta \) becomes unstable.
III. Main Results

In this section, we provide our main results. The threshold for the phase shift \( \alpha \), which ensures stability and depends on the numbers of mediated oscillators and mediators, is presented in the following theorem.

**Theorem 1:** (Threshold of \( \alpha \) for stable remote synchronization) For the dynamics of oscillators (2), the following statements hold:

(i) there exists a unique exponentially stable remote synchronization manifold in \( \mathcal{M}_L \) if
\[
\alpha < \arctan \sqrt{(n + m)/(n - m)}. \tag{3}
\]

(ii) The remote synchronization manifold \( \mathcal{M}_L \) is unstable if
\[
\alpha > \arctan \sqrt{(n + m)/(n - m)}. \tag{4}
\]

Proposition 1 implies that a sufficiently large phase shift \( \alpha \) (or a large time delay, equivalently) prevents stable remote synchronization. It can be seen that \( \arctan \sqrt{(n + m)/(n - m)} \) is monotonically decreasing with respect to \( n \) and monotonically increasing with respect to \( m \). Thus, a larger \( n \) (i.e., more mediated oscillators) results in a narrower range of phase shifts such that an exponentially stable remote synchronization manifold exists. Also, \( \lim_{n \to \infty} \arctan \sqrt{(n + m)/(n - m)} = \pi/4 \) for any given \( m \), which means that exponentially stable remote synchronization always exists regardless of the number of mediated oscillators as long as \( \alpha < \pi/4 \). In contrast, a larger \( m \) (i.e., more mediators) creates a wider range of \( \alpha \) for a given \( n \). In other words, more mediators make remote synchronization more robust against phase shifts.

Before providing the proof of Theorem 1, we present some interesting intermediate results, which will be used to construct the proof.

A. Intermediate Results

Unlike the classic Kuramoto model, e.g., [26], [27], the linearization method cannot be directly used to construct the proof in our case, since the oscillators' frequencies converge to a value that is distinct from the simple average of the natural frequencies [29] due to the presence of the phase shift \( \alpha \). To overcome this problem, we define some new variables.

Let \( x_i = \theta_{i+1} - \frac{1}{n} \sum_{j=1}^{n} \theta_j \) for all \( i \in \mathcal{N}_{n-1} \), and \( y_p = \theta_p - \frac{1}{m} \sum_{j=1}^{m} \theta_j \) for all \( p \in \mathcal{N}_{m} \). Following (2), the time derivatives of these new variables are

\[
\dot{x}_i = \frac{m}{q} \sum_{q=1}^{m} \sin((q - x_i - \alpha) - \frac{n-1}{q} \sum_{j=1}^{m} \sin((j - x_j - \alpha)) + \frac{1}{n} \sum_{j=1}^{n} \sin((j - x_j - \alpha)) - \frac{n}{q} \sum_{j=1}^{n} \sin((j - x_j - \alpha)) - \frac{m}{q} \sum_{j=1}^{m} \sin((j - x_j - \alpha)), \tag{5a}
\]

\[
\dot{y}_p = \frac{n}{j} \sum_{j=1}^{n} \sin((j - y_p - \alpha) + \sin(-y_p - \frac{1}{n} \sum_{j=1}^{n} \sin((j - x_j - \alpha)) - \frac{m}{q} \sum_{j=1}^{m} \sin((j - x_j - \alpha)), \tag{5b}
\]

where \( i \in \mathcal{N}_{n-1} \) and \( p \in \mathcal{N}_{m} \). Denote \( x := [x_1 \ldots x_{n-1}]^\top \in \mathbb{S}^{n-1} \) and \( y := [y_1 \ldots y_m]^\top \in \mathbb{S}^m \).

From Definition 1, a solution \( \theta(t) \) to (2) is remotely synchronized if and only if: 1) \( x = 0 \), and 2) \( \dot{x} = 0 \) and \( \dot{y} = 0 \). Any \((x, y)\) satisfying 1) and 2) is an equilibrium of (5). The following proposition states how the stability of remote synchronization in (2) can be analyzed by studying that of the equilibrium points of (5).

**Proposition 1:** (Connections between remote synchronization in (2) and equilibria of (5)) The equilibria that satisfy \( x = 0 \) of the system (5) in \( \mathbb{S}^{n-1} \) are given by
\[
e_1 = [0_{n-1}, (\pi - c(\alpha) - 2\alpha)]^\top, \tag{6}
\]
\[
e_2 = [0_{n-1}, (\pi + c(\alpha))]^\top, \tag{7}
\]
with
\[
c(\alpha) = -\arctan \left( \frac{(n-m) \sin \alpha + m \sin 3\alpha}{(n+m) \cos \alpha + m \cos 3\alpha} \right), \tag{8}
\]
for any \( p, q \in \mathcal{N}_{m} \). For a given pair of \( p, q \), to ensure (9), either \( y_p = y_q \) or \( y_p + \alpha = \pi - (y_q + \alpha) \) needs to hold. Consequently, at remote synchronization the elements in \( y \) are not necessarily identical, but can be clustered into two groups. Assume that the sizes of these two groups are \( m_1 \) and \( m_2 \), respectively, where \( 0 \leq m_1 \leq m \) and \( m_1 + m_2 = m \). Without loss of generality, let \( y_p = y_p^* \) for \( p = 1, \ldots, m_1 \), and \( y_p = \pi - y_p^* - 2\alpha \) for \( p = m_1 + 1, \ldots, m \). Substituting these \( y_p^* \)'s into the equations \(-n \sin((y_p + \alpha) - \sum_{q=1}^{m_1} \sin(y_q - \alpha) = 0 \) and solving them we obtain two sets of solutions in \( \mathbb{S}^1 \), i.e., 1) \( y_p^* = c(\alpha) \), and 2) \( y_p^* = \pi + c(\alpha) \), where \( c(\alpha) \) is given in (8). Then, (6) and (7) follow subsequently.

Because all the equilibria in (6) and (7) together exhaust all the possible equilibria satisfying \( x = 0 \) of (5) in \( \mathbb{S}^{n-1} \) and each corresponds to a remote synchronization manifold in \( \mathcal{M}_L \), there exists a unique exponentially stable remote synchronization manifold in \( \mathcal{M}_L \) if and only if one of these equilibria is stable, and the remote synchronization manifold \( \mathcal{M}_L \) is unstable if and only if all the equilibria are unstable.

For notational simplicity, let \( s_1 = (n - m_1) \sin \alpha + m_2 \sin 3\alpha \), \( s_2 = (n + m_1) \cos \alpha + m_2 \cos 3\alpha \), and \( S = \ldots \)

The equilibria given in (6) and (7) do not exhaust all the possible equilibria of (5). Other equilibria that do not satisfy \( x = 0 \) may also exist.
\[ \sqrt{s_1^2 + s_2^2}. \] Then, it follows from (8) that
\[
\sin c(\alpha) = -\frac{s_1}{S}, \quad \cos c(\alpha) = \frac{s_2}{S}.
\] (10)

The stability of the equilibria given in (6) and (7) can in the first instance be examined using the Jacobian matrix of (5) evaluated at \( x = 0 \):
\[
J(y) = \begin{bmatrix} R_1(y) & 0 \\ 0 & R_2(y) \end{bmatrix},
\] (11)

where
\[
R_1(y) = -\sum_{q=1}^{m} \cos(y_q - \alpha)I_{n-1},
\]
\[ R_2(y) = -D(y) - C(y), \] (13)

with
\[
C(y) = 1_m^T[\cos(y_1 - \alpha), \ldots, \cos(y_m - \alpha)],
\]
\[
D(y) = \begin{bmatrix} n\cos(y_1 + \alpha) & 0 & \cdots & 0 \\ 0 & n\cos(y_2 + \alpha) & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & n\cos(y_m + \alpha) \end{bmatrix}.
\]

Accordingly, we investigate the eigenvalues of \( J(y) \) at
\[
y = [\cos(c(\alpha) - 2\alpha)I_{m_2}]^T := e^{\varphi}_{1},
\]
\[
y = [\cos(\alpha)I_{m_1},(-\cos(\alpha) - 2\alpha)I_{m_2}]^T := e^{\varphi}_{2},
\] (15)

for all the allowed pairs of \( m_1, m_2 \). Since \( J(y) \) has a block diagonal form, its eigenvalues are composed of those of \( R_1 \) and \( R_2 \). Using this property, we find that some of the equilibria in (6) and (7) are always unstable.

**Proposition 2:** (Unstable equilibria) Let \( m \geq 2 \). Then, all the equilibria in (6) and (7) are unstable for any \( \alpha \in (0, \pi/2) \) and any \( m_1 \) satisfying \( 0 \leq m_1 < m \).

**Proof:** We construct the proof by showing that for any \( m_1 \) satisfying \( 0 \leq m_1 < m \) either \( R_1(y) \) or \( R_2(y) \) has at least one positive eigenvalue no matter whether they are evaluated at \( y = e^{\varphi}_{1} \) or at \( y = e^{\varphi}_{2} \).

First, we consider the case when \( y = e^{\varphi}_{1} \). For any \( i \), it follows that \( C_{ij} = \cos(c(\alpha) - \alpha) \) for \( j = 1, \ldots, m_1 \), \( C_{ij} = -\cos(c(\alpha) + 3\alpha) \) for \( j = m_1 + 1, \ldots, m \), and \( D_{ii} = n\cos(c(\alpha) + \alpha) \) for \( i = 1, \ldots, m_1 \), \( D_{ii} = -n\cos(c(\alpha) + \alpha) \) for \( i = m_1 + 1, \ldots, m \). For notational simplicity, let \( a = \cos(c(\alpha) + \alpha) \), \( b = \cos(c(\alpha) - \alpha) \), and \( c = -\cos(c(\alpha) + 3\alpha) \), and then \( R_1(y), D, \) and \( C \) can be rewritten as
\[
R_1(y) = -((m-1)b+c)I_{n-1},
\]
\[
C = 1_m^T[1_{m_1}, e^{\varphi}_{1}]^T, \quad D = [na1_{m_1}, 0, -na1_{m_2}].
\] (17)

We then show that \( D + C \) has a negative eigenvalue in both the following two cases: a) when \( m_1 \leq m - 2 \); b) when \( m_1 = m - 1 \). We start with the case a). Let \( v_1 := [0^T_{n-1}, -1, 0]^T \), and then \( (D + C)v_1 = Dv_1 + 0 = -nav_1 \), which means that \(-na\) is an eigenvalue of \( D + C \). As a consequence, \( J(y) \) has at least one positive eigenvalue. We then study the case b), and show that either \( R_1(y) \) or \( R_2(y) \) has a positive eigenvalue. From (16), all the eigenvalues of \( R_1(y) \) are identical and equal to \(-((m-1)b+c)\). To ensure that all the eigenvalues of \( J(y) \) have negative real parts, \(((m-1)b+c) > 0\) needs to hold. We then prove that even when \(((m-1)b+c) > 0\) the matrix \( R_2(y) \) still has a positive eigenvalue. We prove that fact by showing that there is \( \phi \) such that \( v_2 := [1^T_{m-1}, \phi]^T \) is the eigenvector of \( D + C \) and it is associated with a negative eigenvalue (denoted by \( \lambda \)). Let \((D + C)v_2 = \lambda v_2 \), and we obtain \( [na1_{m-1} - na\phi] + (m-1)b + c\phi \) \( I_m \alpha \) \( = \lambda \alpha I_{m-1} \), \( \phi \), from which we have the following two equations
\[
a + (m-1)b + c\phi = \lambda, -na\phi + (m-1)b + c\phi = \lambda. \] (18)

We then show that there is a pair of solutions \( \phi \) and \( \lambda \) to the above equations, satisfying \( \lambda < 0 \). Canceling \( \phi \) in the above equations we obtain the equation of \( \lambda \) as follows
\[
\lambda^2 - ((m-1)b+c)\lambda - na((m-1)b+na-c) = 0.
\]

The solutions to the above quadratic equation are \( \lambda_1 = \frac{2m+1}2 - \sqrt{\frac{2m+1}2 + 4na} \) and \( \lambda_2 = \frac{2m+1}2 + \sqrt{\frac{2m+1}2 + 4na} \). Since \( n_2 > 0 \) from Proposition 3 in the Appendix and \( \alpha > 0 \) from Lemma 1 in the Appendix, it is not hard to see that \( \lambda_2 < 0 \). Substituting \( \lambda_2 \) into (18), one can compute the solution \( \phi \), which means that \( \lambda_2 \) is an eigenvalue of \( D + C \) that is associated with the eigenvector \( v_2 := [1^T_{m-1}, \phi]^T \). Therefore, we have proven that \( J(y) \) evaluated at \( y = e^{\varphi}_{1} \) has at least one positive eigenvalue for any \( m_1 < m \).

Finally, following similar steps as above one can prove that \( J(y) \) evaluated at \( y = e^{\varphi}_{2} \) also has at least one positive eigenvalue, which completes the proof.

**B. Proof of Theorem 1**

**Proof of Theorem 1:** From Proposition 1, we construct the proof by showing that: I) when \( m_1 < m \) (which implies \( m \geq 2 \), since by definition \( m_1 = m \) when \( m = 1 \)), all the equilibria in (6) and (7) are unstable for any \( \alpha \); and II) when \( m_1 = m \) for any \( m \geq 1 \), \( \epsilon_1 \) is exponentially stable under (3) and unstable under (4), and \( \epsilon_2 \) is unstable for any \( \alpha \).

First, the proof of I) follows directly from Proposition 2.

Second, when \( m_1 = m = \alpha \) in (8) becomes
\[
c(\alpha) = -\arctan \left( \frac{n \tan \alpha}{m + \sin \alpha} \right), \quad \epsilon_1, \epsilon_2 \text{ become } e_1 = \left[ 0^T_{m-1}, c(\alpha)1^T_m \right]^T \text{ and } e_2 = \left[ 0^T_{m-1}, (\pi - c(\alpha))1^T_m \right]^T \text{, respectively.}
\]

We prove II) by showing the following two facts: a) \( J(y) \), evaluated at \( y = e^{\varphi}_{1} = c(\alpha)1^T_m \), is Hurwitz under (3), and has positive eigenvalues under (4); b) \( J(y) \), evaluated at \( y = e^{\varphi}_{2} = (\pi - c(\alpha))1^T_m \), has positive eigenvalues for any \( \alpha \).

To prove a), we investigate the eigenvalues of \( R_1(y) \) and \( R_2(y) \) at \( y = c(\alpha)1^T_m \). It follows from (12) and (13) that
\[
R_1(y) = -m \cos(c(\alpha) - \alpha)I_{n-1},
\]
\[
R_2(y) = -m \cos(c(\alpha) + \alpha)I_m - \cos(c(\alpha) - \alpha)1^T_m 1_m.
\] (20)

All the eigenvalues of \( R_1 \) are \(-m \cos(c(\alpha) - \alpha)\). Moreover,
\[
\cos(c(\alpha) - \alpha) = \frac{1}{2} \left( (n + m) \cos^2 \alpha - (n - m) \sin^2 \alpha \right),
\] (21)

The right-hand side of (21) is positive (negative, respectively) under (3) (4), respectively, which means that the
The frequencies of the mediated oscillators converge to a very large value equal to \( \eta_1 \) for any \( \alpha \). Consequently, given \( \eta_1, \eta_2 \in \mathbb{R} \), the matrix \( \eta_1 I_m + \eta_2 1_m 1_m^\top \) has \( m - 1 \) eigenvalues that equal 0 and one eigenvalue that is \( m \). Moreover, it appears that \( \alpha \in \mathcal{N}_m \); then evidently \( \mu_i = -n \cos(\alpha) + \alpha \) for \( i = 1, \ldots, m - 1 \), and \( \mu_m = -n \cos(\alpha + \alpha) - m \cos(\alpha - \alpha) \). For \( s_1 \) and \( s_2 \) given in (10), \( s_1 = (n - m) \sin \alpha \) and \( s_2 = (n + m) \cos \alpha \), and then \( n \cos(\alpha + \alpha) + m \cos(\alpha - \alpha) = \frac{1}{2} ((n + m)^2 \cos^2 \alpha + (n - m)^2 \sin^2 \alpha) \), which is positive for any \( \alpha \). Also, \( \cos(\alpha + \alpha) > 0 \) for any \( \alpha \) from Lemma 1 in the Appendix, and thus the eigenvalues of \( R_2 \) are negative for any \( \alpha \). Overall, \( J(y) \), evaluated at \( y = e_1 \), is Hurwitz under (3), and has positive eigenvalues under (4).

We finally prove b). At \( y = (\pi + c(\alpha))1_m \), following similar steps as above, the eigenvalues of \( R_2(y) \) are \( \mu_1 = \cdots = \mu_{m-1} = n \cos(\alpha) + \alpha, \mu_m = n \cos(\alpha + \alpha) + m \cos(\alpha - \alpha) \). Then, all the eigenvalues of \( R_2(y) \) are positive for any \( \alpha \), which subsequently means that \( J(y) \) has positive eigenvalues for any \( \alpha \). The proof is complete.

We have proven in Theorem 1 that exponentially stable remote synchronization is possible only when \( m_1 = m \) (under a certain condition on \( \alpha \)). Note that \( m_1 = m \) implies that the mediators have an identical phase. Therefore, to ensure the exponential stability of phase-locked remote synchronization, the mediators themselves have to be synchronized, but their phases usually differ from the mediated oscillators. In the phase-unlocked case, however, the mediators can be incoherent while still guaranteeing stable remote synchronization, which will be shown numerically in the next section. Finally, the equilibrium \( e_1 \), when \( m_1 = m \), is the only equilibrium which can be exponentially stable, and its stability depends only on \( R_1(y) \) in (19) as \( R_2(y) \) is always Hurwitz. The calculation in (21) shows that, as \( \alpha \) approaches \( \arctan \sqrt{(n + m)/2} \) from below, \( R_1(y) \rightarrow 0 \), so that the system has a progressively smaller degree of stability.

**IV. SIMULATION RESULTS**

In this section, we present a set of illustrative simulations that are beyond our theoretical results, and provide some interesting observations.

First, we investigate the situation when \( m \geq n \). Phase differences in the case of \( m = n = 3 \) are plotted in Fig. 2(a). It appears that complete (not just remote) synchronization can occur for any \( \alpha \in (0, \pi/2) \), since it is observed for a very large \( \alpha \), i.e., \( \pi/2 - 0.1 \). When \( n = 3 \) and \( m = 4 \), it can be observed from Fig. 2(b) that phase synchronization of the mediated oscillators always takes place even when the phase shift \( \alpha \) is large (\( \alpha_1 = 1.35, \alpha_2 = 1.4 \)). However, the frequencies of the mediated oscillators’ converge to a dynamically changing value that is quite distinct from those of the mediators (see Fig. 2(c)). This implies that phase-unlocked remote synchronization has occurred. Moreover, the frequencies of the mediators stay distinct from one another, which implies that their phases also remain distinct. This phenomenon is known as a Chimera state [30], since synchronization and desynchronization coexist in the same network. Interestingly, simulation results confirm the occurrence of remote synchronization for any \( \alpha \). We conjecture that the mediators’ quantitative advantage creates a powerful structure, which can eliminate the effect of any phase shift or time delay, making remote synchronization always stable.

Clearly, complete bipartite networks are a special class of networks. Real networks, such as brain networks, are certainly not bipartite. Yet, bipartite components can be found in complex networks including brain networks, and reasonably might play a role in enforcing synchronization. Then, we consider a network with a bipartite component in Fig. 2(d). It is shown in Fig. 2(e) that the oscillators 1 and 2, mediated by \( r_1, r_2, r_3 \), gradually become synchronized for a wide range of phase shift \( \alpha \) (the cases when \( \alpha \) equals \( \beta_1 = 0.6, \beta_2 = 0.8, \beta_3 = 1, \) and \( \beta_4 = 1.2 \) are plotted). However, if we reduce the number of mediators by removing the red dashed edges in Fig. 2(d), we find that remote synchronization cannot appear anymore even when the phase shift \( \alpha \) is as small as 0.9. From the simulations, we confirm that a bipartite component in a network plays an important role in ensuring stable remote synchronization. Moreover, more mediators make remote synchronization more robust against phase shifts (or time delays), which is similar to our theoretical findings in bipartite networks.

**V. CONCLUSION**

Cortical regions without apparent neuronal links exhibit synchronized behaviors, and the common neighbors that they share seem to play crucial roles in giving rise to this phenomenon. Motivated by these empirical observations, we have analytically studied mediated remote synchronization...
of Kuramoto-Sakaguchi oscillators coupled by bipartite networks. A larger number of mediators has been shown to make remote synchronization more robust to phase shifts or time delays. Simulation results confirm that this finding also applies to more complex networks with bipartite components. Moreover, remote synchronization seems to be stable for any phase shift if there are more mediators than the mediated oscillators in a bipartite network. This aspect is left as the subject of future investigation.

APPENDIX

Lemma 1: Let \( c(\alpha) \) be defined in (8), then \( \cos(c(\alpha) + \alpha) > 0 \) for any \( m_1 \) and any \( \alpha \in (0, \pi/2) \).

Proof: There holds that \( \cos(c(\alpha) + \alpha) = \cos(c(\alpha)) \cos(\alpha) - \sin(c(\alpha)) \sin(\alpha) \). Substituting (10) into the right-hand side of this inequality we can compute

\[
\cos(c(\alpha) + \alpha) = \frac{1}{2} (n - m + 2m \cos^2 \alpha),
\]

where the last equality has used the double-angle formula \( \cos 2\alpha = 2 \cos^2 \alpha - 1 \) and the equality \( m_1 + m_2 = n \). Then, \( \cos(c(\alpha) + \alpha) > 0 \) for any \( \alpha \) and \( m_1 \) since \( m < n \) by hypothesis.

Proposition 3: Let \( a = \cos(c(\alpha) + \alpha) \), \( b = \cos(c(\alpha)) \), and \( c = -\cos(c(\alpha) + 3\alpha) \), where \( c(\alpha) \) is given by (8). Given \( m \geq 2 \), suppose \( m_1 = m - 1 \), then \( (m - 1)b + na - c > 0 \) for any \( \alpha \in (0, \pi/2) \).

Proof: Substituting (8) into \((m - 1)b + na - c\) and after some algebra one can obtain \( S \) of \((m - 1)b + na - c \). The solution is given by

\[
S \geq 16 \cos^4 \alpha + 8 \cos^2 \alpha + 1 > 0.
\]

Therefore, \( S \) is always positive, and it follows naturally that \( (m - 1)b + na - c > 0 \).

REFERENCES

[1] J. M. Palva, S. Palva, and K. Kaila, “Phase synchrony among neuronal oscillations in the human cortex,” *Journal of Neuroscience*, vol. 25, no. 15, pp. 3962–3972, 2005.

[2] T. Womelsdorf, J.-M. Schoffelen, R. Oostenveld, W. Singer, R. Desimone, A. K. Engel, and P. Fries, “Modulation of neuronal interactions through neuronal synchronization,” *Science*, vol. 316, no. 5831, pp. 1609–1612, 2007.

[3] P. Fries, “A mechanism for cognitive dynamics: Neuronal communication through neuronal coherence,” *Trends in Cognitive Sciences*, vol. 9, no. 10, pp. 474–480, 2005.

[4] A. Palmigiano, T. Geisel, F. Wolf, and D. Battaglia, “Flexible information routing by transient synchrony,” *Nature Neuroscience*, vol. 20, no. 7, pp. 1014, 2017.

[5] C. J. Honey, R. Kotter, M. Breakspear, and O. Sporns, “Network structure of cerebral cortex shapes functional connectivity on multiple time scales,” *Proceedings of the National Academy of Sciences*, vol. 104, no. 24, pp. 10 240–10 245, 2007.

[6] E. Bullmore and O. Sporns, “Complex brain networks: Graph theoretical analysis of structural and functional systems,” *Nature Reviews Neuroscience*, vol. 10, no. 3, pp. 186–198, 2009.

[7] F. Varela, J.-P. Lachaux, E. Rodriguez, and J. Martinerie, “The brainweb: Phase synchronization and large-scale integration,” *Nature Reviews Neuroscience*, vol. 2, no. 4, pp. 229–239, 2001.

[8] Y. Mochizuki, T. Onaga, H. Shimazaki, and et al., “Similarity in neuronal firing regimes across mammalian species,” *Journal of Neuroscience*, vol. 36, no. 21, pp. 5736–5747, 2016.

[9] R. F. Betzel, J. D. Medaglia, A. E. Kahn, J. Soffer, D. R. Schonhaut, and D. S. Bassett, “Structural, geometric and genetic factors predict interregional brain connectivity patterns probed by electrocorticography,” *Nature Biomedical Engineering*, vol. 3, no. 11, pp. 902–916, 2019.

[10] V. Nicosia, M. Valencia, M. Chavez, A. Diaz-Guilera, and V. Latora, “Remote synchronization reveals network symmetries and functional modules,” *Physical Review Letters*, vol. 110, no. 17, p. 174102, 2013.

[11] R. Vicente, L. L. Gollo, C. R. Mirasso, I. Fischer, and G. Pipa, “Dynamical relaying can yield zero time lag neuronal synchrony despite long conduction delays,” *Proceedings of the National Academy of Sciences*, vol. 105, no. 44, pp. 17 157–17 162, 2008.

[12] V. Vukovárová and P. Hövel, “Functional connectivity of distant cortical regions: Role of remote synchronization and symmetry in interactions,” *NeuroImage*, vol. 97, pp. 1–8, 2014.

[13] L. Zhang, A. E. Motter, and T. Nishikawa, “Incoherence-mediated remote synchronization,” *Physical Review Letters*, vol. 118, p. 174102, 2017.

[14] F. Dörfler and F. Bullo, “Synchronization in complex networks of phase oscillators: A survey,” *Automatica*, vol. 50, no. 6, pp. 1539–1564, 2014.

[15] T. Menara, G. Baggio, D. S. Bassett, and F. Pasqualetti, “Stability conditions for cluster synchronization in networks of heterogeneous Kuramoto oscillators,” *IEEE Transactions on Control of Network Systems*, 2019, in press.

[16] L. M. Pecora, F. Sorrentino, A. M. Hagerstrom, T. E. Murphy, and R. Roy, “Cluster synchronization and isolated desynchronization in complex networks with symmetries,” *Nature Communications*, vol. 5, no. 1, pp. 1–8, 2014.

[17] F. Sorrentino, L. M. Pecora, A. M. Hagerstrom, T. E. Murphy, and R. Roy, “Complete characterization of the stability of cluster synchronization in complex dynamical networks,” *Science Advances*, vol. 2, no. 4, 2016.

[18] Y. Qin, Y. Kawano, O. Portoles, and M. Cao, “Partial phase cohesiveness in networks of Kuramoto oscillator networks,” arXiv preprint arXiv:1906.01065, 2019.

[19] A. Bergner, M. Frasca, G. Sciuto, A. Buscarino, E. J. Ngamga, L. Fortuna, and J. Kurths, “Remote synchronization in star networks,” *Physical Review E*, vol. 85, no. 2, p. 026208, 2012.

[20] L. V. Gambuzza, A. Cardillo, A. Fiasconaro, L. Fortuna, J. Gómez-Gardenes, and M. Frasca, “Analysis of remote synchronization in complex networks,” *Chaos: An Interdisciplinary Journal of Nonlinear Science*, vol. 23, no. 4, pp. 043103, 2013.

[21] Y. Qin, Y. Kawano, and M. Cao, “Stability of remote synchronization in star networks of Kuramoto oscillators,” in *IEEE Conf. on Decision and Control*, Dec 2018, pp. 5209–5214.

[22] H. Sakaguchi and Y. Kuramoto, “A soluble active rotator model showing phase transitions via mutual entertainment,” *Progress of Theoretical Physics*, vol. 76, no. 3, pp. 576–581, 1986.

[23] Y. Kuramoto, “Self-entrainment of a population of coupled non-linear oscillators,” in *International Symposium on Mathematical Problems in Theoretical Physics*, Berlin, Heidelberg, 1975, pp. 420–422.

[24] F. C. Hoppensteadt and E. M. Izhikevich, *Weakly Connected Neural Networks*. Springer Science & Business Media, 2012.

[25] N. Punetha, S. R. Ujjwal, F. M. Atay, and R. Ramaswamy, “Delay-induced remote synchronization in bipartite networks of phase oscillators,” *Physical Review E*, vol. 91, no. 2, p. 022922, 2015.

[26] S. Jalafpour and F. Bullo, “Synchronization of Kuramoto oscillators via cutset projections,” *IEEE Transactions on Automatic Control*, vol. 64, no. 7, pp. 2830–2844, 2018.

[27] F. Dörfler, M. Chertkov, and F. Bullo, “Synchronization in complex oscillator networks and smart grids,” *Proceedings of the National Academy of Sciences*, vol. 110, no. 6, pp. 2005–2010, 2013.

[28] Y. Qin, Y. Kawano, B. D. O. Anderson, and M. Cao, “Partial exponential stability analysis of slow-fast systems via periodic averaging,” arXiv preprint arXiv:1910.07465, 2019.

[29] E. Monthus, J. Kurths, and B. Blasius, “Synchronization of two interacting populations of oscillators,” *Physical Review E*, vol. 70, no. 5, pp. 056125, 2004.

[30] M. J. Panaggio and D. M. Abrams, “Chimera states: Coexistence of coherence and incoherence in networks of coupled oscillators,” *Nonlinearity*, vol. 28, no. 3, pp. R67–R87, 2015.