EMERGENT BEHAVIORS OF HIGH-DIMENSIONAL KURAMOTO MODELS ON STIEFEL MANIFOLDS

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Abstract. We study emergent asymptotic dynamics for the first and second-order high-dimensional Kuramoto models on Stiefel manifolds which extend the previous consensus models on Riemannian manifolds including several matrix Lie groups. For the first-order consensus model on the Stiefel manifold proposed in [Markdahl et al, 2018], we show that the homogeneous ensemble relaxes the complete consensus state exponentially fast. On the other hand for a heterogeneous ensemble, we provide a sufficient condition leading to the phase-locked state in which relative distances between two states converge to definite values in a large coupling strength regime. We also propose a second-order extension of the first-order one by adding an inertial effect, and study emergent behaviors using Lyapunov functionals such as an energy functional and an averaged distance functional.

1. Introduction

Collective behaviors of biological and chemical oscillators have been widely studied not only in applied mathematics but also in other scientific disciplines, for instance, flocking of drones [30, 46] and passivity-based distributed optimization [11, 29, 51] in control theory, social dynamics [2, 4, 22, 44] and swarming behavior in quantitative biology [5, 14, 15, 23, 50]. Despite of its crucial roles in biological processes, the mathematical study of such collective dynamics has been started only after the seminal work of Winfree [61] and Kuramoto [33, 34] a half century ago. Among other models describing collective oscillatory behavior, to name a few, the Cucker-Smale model [13], the Kuramoto model [12, 33, 34] and the Vicsek model [58], our main interest lies in consensus models on Riemannian manifolds. So far, there has been much available literature dealing with emergent dynamics on Riemannian manifolds, for instance, on the unit sphere \( S^{d-1} \) in [45, 62], on the hyperboloid \( H^{d-1} \) in [49] and on the matrix Lie groups including special orthogonal group \( \text{SO}(d) \) in [16, 53, 56], the unitary group \( \text{U}(d) \) in [27, 28, 39], the Lohe group in [26] and even for the quaternions \( \text{H}_1 \) in [17]. In particular, we are concerned with the Stiefel manifold [55] in \( \mathbb{R}^{n \times p} \) for \( p \leq n \) consisting of all rectangular matrices satisfying the relations:

\[
\text{St}(p, n) := \{ S \in M_{n,p}(\mathbb{R}) : S^\top S = I_p \}, \quad \| S \|_F^2 := \text{tr}(S^\top S) = p,
\]

where \( S^\top \) is defined as the transpose of a matrix \( S \) and \( M_{n,p}(\mathbb{R}) \) denotes the set of all \( n \times p \) matrices with real entries.

Note that the unit sphere and (special) orthogonal group can be recovered from the specific choices of \((p, n)\). Thus, the Stiefel manifold is a manifold including \( S^d \), \( \text{SO}(d) \) and...
O(d) as special cases. We will briefly review basic properties of the Stiefel manifold in Section 2.1. It is worthwhile to mention that an optimization problem \( \min_{S \in \text{St}(p,n)} f(S) \) for an objective function \( f : \text{St}(p,n) \to \mathbb{R} \) has been extensively studied due to its computational difficulty and broad applications, e.g., the linear eigenvalue problem \([21, 59]\), the nearest low-rank correlation matrix problem \([35]\), singular value decomposition \([38, 54]\), Riemannian optimization \([63]\) and applications to computer vision \([40, 57]\). We refer the reader to \([18, 52, 60]\) and reference therein for introductions to optimization problem on the Stiefel manifold and applications.

In this paper, we mainly consider the consensus model in \([41, 42]\) in which particles interact with neighboring ones on the Stiefel manifold and for the consistency of the paper, we briefly introduce the model below. Let \((\text{St}(p,n), \| \cdot \|)\) be the Stiefel manifold canonically embedded into the Euclidean space \(\mathbb{R}^{n \times p}\) with its Frobenius (or Euclidean) norm and \(S := (S_1, \cdots, S_N)\) denote state ensemble of particles. Next, we define the potential function \(V\) as the total misfit functional with a symmetric connectivity matrix \(A := (a_{ik})\):

\[
V(S) := \kappa N^2 \sum_{i,k=1}^{N} a_{ik} \|S_i - S_k\|_F^2, \quad a_{ik} = a_{ki} > 0.
\]

Then, the gradient flow with the potential function \(V\) reads as

\[
\dot{S}_i = -\nabla_{S_i} V = \frac{\kappa}{N} \sum_{k=1}^{N} a_{ik} \left[ S_k - \frac{1}{2} (S_i S_i^T S_k + S_k S_i^T S_i) \right], \quad i = 1, \cdots, N. \tag{1.1}
\]

See Section 2.2 for detailed description. We finally add generalized natural frequencies \(\Xi_i \in \mathfrak{so}(p)\) into \((1.1)\) to find the desired model so that the Stiefel manifold is still positively invariant:

\[
\dot{S}_i = S_i \Xi_i + \frac{\kappa}{N} \sum_{k=1}^{N} a_{ik} \left[ S_k - \frac{1}{2} (S_i S_i^T S_k + S_k S_i^T S_i) \right], \quad i = 1, \cdots, N. \tag{1.2}
\]

Since the Stiefel manifold is compact, the global existence of a unique solution to \((1.2)\) directly follows from the standard Cauchy-Lipschitz theory. Detailed description will be presented in Section 2.1.

For the perspective of optimization, a gradient flow \((1.1)\) with a total distance as a potential can be regarded an optimization problem of a given target (or objective) function defined on the Stiefel manifold. Thus, our model with a consensus estimate provides a method which tackles an optimization problem on the Stiefel manifold by means of dynamical systems approach. It should be mentioned that such optimization problems on manifolds have extensively studied in literature \([4, 7, 37, 48, 52]\). In particular, consensus-based optimization (CBO for brevity) algorithm toward a global optimization \([6, 19, 20, 22, 47]\) has been recently proposed (see \([31]\) for CBO on the Stiefel manifold). On the other hand for a control (or consensus protocol) perspective, it is mentioned in \([43]\) that \textit{ad-hoc} control algorithm would be employed in a specific situation. In this manner, one can consider \((1.1)\) as a control problem by regarding the network structure \(a_{ik}\) as an external control (or parameter) to obtain a desired pattern formation (see \([36]\) on the unit sphere and \textit{an Olympic ring} in \([10]\) for a distributed control approach for the Cucker-Smale flocking model \([13]\)). For instance, if we set \(a_{ik}\) to be negative (or competitive), one can expect deployment where all agents tend to splay state, whereas complete consensus would be achieved for a positive
More precisely, we choose all-to-all network topology so that emergence of complete consensus is our primary concern.

We analyze system (1.2) with all-to-all network topology and left-translation invariance. More precisely, we choose all-to-all network \( a_{ij} = \kappa/N \) where \( \kappa \) denotes the coupling strength so that all particles communicate with neighbors through the same weight. In addition, we do not consider the effect of \( \Omega_i \) by setting \( \Omega_i \equiv O \) due to the left-translation property (see Lemma 2.2 for details). Then, our first-order consensus model is governed by the following Cauchy problem:

\[
\begin{aligned}
\dot{S}_i &= S_i \Xi_i + \frac{\kappa}{N} \sum_{k=1}^{N} \left[ S_k - \frac{1}{2} (S_i S_i^T S_k + S_i S_k^T S_i) \right], \quad t > 0, \\
S_i(0) &= S_{i}^{in} \in \text{St}(p, n), \quad i = 1, \cdots, N.
\end{aligned}
\]  

(1.3)

Although the gradient flow structure guarantees that the Stiefel manifold is positively invariant along the flow (1.2), we provide its alternative and direct proof in Lemma 2.1.

Next, we turn to the second-order extension of the first-order model (1.3). In [25], the (first-order) Lohe matrix model on the unitary group has been extended to a second-order variant along the flow (1.2), we provide its alternative and direct proof in Lemma 2.1.

Next, we turn to the second-order extension of the first-order model (1.3). In [25], the Lohe matrix model on the unitary group has been extended to a second-order variant along the flow (1.2), we provide its alternative and direct proof in Lemma 2.1.

Similarly, we can also extend the first-order model (1.3) into a second-order one whose dynamics is governed by the following Cauchy problem:

\[
\begin{aligned}
m \dot{S}_i &= -m S_i S_i^T \dot{S}_i - \gamma \dot{S}_i + S_i \Xi_i + \frac{m}{\gamma}(2 \dot{S}_i - S_i S_i^T \dot{S}_i), \\
&\quad + \frac{\kappa}{N} \sum_{k=1}^{N} \left[ S_k - \frac{1}{2} (S_i S_i^T S_k + S_i S_k^T S_i) \right], \quad t > 0, \\
S_i(0) &= S_{i}^{in} \in \text{St}(p, n), \quad \dot{S}_{i}^{t, in}, S_{i}^{t, in} = O, \quad i = 1, \cdots, N.
\end{aligned}
\]  

(1.4)

where \( m \) and \( \gamma \) represent mass and friction, respectively. Although it seems that the model looks quite complicated, we show the the Stiefel manifold is still positively invariant along system (1.4). Of course, if we turn off the inertial effect, that is, \( m = 0 \), then the first-order model (1.3) can be recovered from (1.4) straightforwardly. We also note that the Cauchy problems (1.3) and (1.4) have a unique global solution due to the compactness of Stiefel manifold and standard Cauchy-Lipschitz theory. Here, a global solution is referred as a solution which exists for all time (or globally). In other words, a solution does not blow up in finite time. For more detailed description of (1.4), we refer the reader to Section 2.2.

Next, we recall several concepts for consensus as follows:

**Definition 1.1.** Let \( \mathcal{S} = (S_1, \cdots, S_N) \) be a global solution to (1.3) or (1.4).

1. **System** (1.3) or (1.4) exhibits complete consensus if the following convergence holds:

\[
\lim_{t \to \infty} \| S_i(t) - S_j(t) \|_F = 0, \quad \text{for all } i, j = 1, \cdots, N.
\]

2. **System** (1.3) or (1.4) exhibits practical consensus if the following convergence holds:

\[
\lim_{\kappa \to \infty} \limsup_{t \to \infty} \| S_i(t) - S_j(t) \|_F = 0, \quad \text{for all } i, j = 1, \cdots, N.
\]

3. A global solution to system (1.3) or (1.4) tends to a phase-locked state if the following relation holds:

\[
\lim_{t \to \infty} S_i^T S_j \text{ exists for all } i, j = 1, \cdots, N.
\]
The main results of this paper deal with the emergent collective behaviors for the first-order model (1.3) and the second-order model (1.4). First, we consider the first-order model (1.3) with a homogeneous ensemble. The corresponding proof can be found in Section 3.1.

**Theorem 1.1.** Suppose that initial data and system parameters satisfy
\[ D(S^{\text{in}}) < \sqrt{2}, \quad \Xi \equiv O, \quad a_{ik} : \text{undirected and connected graph}, \]
and let \( S \) be a global solution to (1.2). Then, system (1.2) exhibits complete consensus exponentially fast.

The results in [41, 42] deal with almost global consensus without any explicit decay estimate. On the other hand, for a heterogeneous ensemble, we establish the emergence of the locked state exponentially fast in a large coupling regime.

**Theorem 1.2.** Suppose that the coupling strength and initial data satisfy (3.11) and let \( S \) be a global solution to (1.2). Then, \( S \) tends to a locked state.

For a detailed initial framework and proof, we refer to Section 3.2.

Next, we turn to the second-order model (1.4). As in the first-order one, we consider both homogeneous and heterogeneous ensembles. For the desired results, we first derive an energy estimate (Proposition 4.1):
\[
\begin{align*}
\frac{d}{dt} \left( m \frac{N}{N} \sum_{i=1}^{N} \| \dot{S}_i \|^2_F + \frac{\kappa}{2N^2} \sum_{i,j=1}^{N} \| S_i - S_j \|^2_F \right) \\
= -2\gamma \frac{N}{N} \sum_{i=1}^{N} \| \dot{S}_i \|^2_F + \frac{1}{N} \sum_{i=1}^{N} \text{tr}(\dot{S}_i^\top S_i \Xi_i - \Xi_i \dot{S}_i^\top \dot{S}_i), \quad t > 0.
\end{align*}
\] (1.5)

In what follows, we assume that the network topology satisfies
\[
a_m := \min a_{ik}, \quad a_M := \max a_{ik}, \quad d(A) := \max |a_{ik} - a_{jk}|,
\]
\[
\frac{1}{N} \sum_{k=1}^{N} a_{ik} \equiv \xi, \quad i = 1, \cdots, N, \quad 0 < \Lambda := a_m - \frac{N - 1}{N} (a_M + d(A)) < 8\rho a_M^2.
\]

Furthermore, for the averaged relative distances \( \mathcal{G} := \frac{1}{N^2} \sum_{i,j=1}^{N} \| S_i - S_j \|^2_F \), we derive a second-order differential inequality (Lemma 4.3) for \( \mathcal{G} \):
\[
m\ddot{\mathcal{G}} + \gamma \dot{\mathcal{G}} + 2\kappa \xi \mathcal{G} \leq 16mD(\dot{S})^2 + 8\| \Xi \|_\infty + \frac{16m\sqrt{p}\| \Xi \|_\infty}{\gamma} D(\dot{S}),
\] (1.6)

where \( D(\dot{S}) \) and \( \| \Xi \|_\infty \) are defined as follows:
\[
D(\dot{S}) := \max_{1 \leq i \leq N} \| \dot{S}_i \|_F, \quad \| \Xi \|_\infty := \max_{1 \leq i \leq N} \| \Xi_i \|_F.
\]

Based on two estimates (1.5) and (1.6), for a homogeneous ensemble, we show that the complete consensus occurs for some admissible initial data.
Theorem 1.3. Suppose that system parameters and initial data satisfy
\[ m > 0, \quad \gamma > 0, \quad \kappa > 0, \quad \Xi_i \equiv O \quad \text{for} \quad i = 1, \cdots, N, \]
\[ \mathcal{E}(0) = \frac{m}{N} \sum_{i=1}^{N} \| \dot{S}_i^0 \|^2_F + \frac{\kappa}{2N^2} \sum_{i,j=1}^{N} a_{ij} \| S_i^0 - S_j^0 \|^2_F < \infty, \]
and let \( S \) be a global solution to (1.4). Then, we have
\[ \lim_{t \to \infty} \| \dot{S}_i(t) \|_F = 0 \quad \text{for} \quad i = 1, \cdots, N. \]
Moreover, system (1.4) exhibits the complete consensus:
\[ \lim_{t \to \infty} \mathcal{G}(t) = 0. \]

We refer the reader to Section 4.1 for the proof. In contrast, for a heterogeneous ensemble, we assume that the inertia and the coupling satisfy the following relation:
\[ mk\kappa^{1+\eta} = \mathcal{O}(1) \quad \text{for some} \quad \eta > 0. \]
Then, under this setting, we arrive at the following result.

Theorem 1.4. Suppose that system parameters and initial data satisfy
\[ D(\dot{S}^0) < \frac{1}{\gamma}(\| \Xi \|_{\infty} + \kappa a_M \sqrt{p}), \quad m = \frac{m_0}{\kappa^{1+\eta}}, \tag{1.7} \]
and let \( S \) be a global solution to (1.4). Then, system (1.4) exhibits practical consensus:
\[ \lim_{\kappa \to \infty} \limsup_{t \to \infty} \mathcal{G}(t) = 0. \]

The proof can be found in Section 4.2.

The rest of the paper is organized as follows. In Section 2, we briefly discuss properties of the Stiefel manifold to be used later and provide descriptions of the first-order and second-order models. In Section 3, we present proofs of Theorem 1.1 and Theorem 1.2 which deal with the first-order model. In Section 4, rigorous justification of Theorem 1.3 and Theorem 1.4 for the second-order model is provided. Finally, Section 5 is devoted to a brief summary of our main results and discussion for a future work.

Notation: We denote by \( M_{n,p}(\mathbb{R}) \) as the set of all \( n \times p \) real matrices and for notational simplicity, we set \( M_n(\mathbb{R}) := M_{n,n}(\mathbb{R}) \). In addition, \( O \) is the zero matrix regardless of its size.

2. Preliminaries

In this section, we briefly discuss the Stiefel manifold and present detailed description and properties of the first-order and second-order consensus models on the Stiefel manifold to be used later in later sections.
2.1. The Stiefel manifold. We define Stiefel manifold and Frobenius norm:

\[ \text{St}(p, n) := \{ S \in M_{n,p}(\mathbb{R}) : S^\top S = I_p \}, \quad ||S||_F^2 := \text{tr}(S^\top S) = p \quad \text{for } S \in \text{St}(p, n). \]

Alternatively, it can be defined as the set of all \( p \)-tuples \( (x_1, \cdots, x_p) \) of orthonormal vectors in \( \mathbb{R}^n \) or it is isomorphic to a homogeneous space:

\[ \text{St}(p, n) \simeq O(n)/O(n-p). \]

In addition, if \( p \) is strictly less than \( n \), then one also finds

\[ \text{St}(p, n) \simeq SO(n)/SO(n-p). \]

Thus, the Stiefel manifold \( \text{St}(p, n) \) is a compact set whose dimension is \( pn - p(p+1)/2 \). Furthermore, it is well known that \( \text{St}(p, n) \) reduces to several well-known manifolds, for instance,

\[ \text{St}(1, n) = \mathbb{S}^{n-1} \subseteq \mathbb{R}^n, \quad \text{St}(n-1, n) = SO(n), \quad \text{St}(n, n) = O(n). \]  

(2.1)

We set \( \frak{o}(n) \) to be the special orthogonal Lie algebra associated with \( SO(n) \). Then, we define two maps \( \text{skew} : M_n(\mathbb{R}) \to \frak{o}(n) \) and \( \text{sym} : M_n(\mathbb{R}) \to \frak{o}(n)^\perp \) as

\[ \text{skew}(X) := \frac{1}{2}(X - X^\top), \quad \text{sym}(X) := \frac{1}{2}(X + X^\top). \]

Then, it is easy to see that the tangent space and the normal space of \( \text{St}(p, n) \) at a point \( S \) are defined by

\[ T_S\text{St}(p, n) := \{ A \in M_{n,p}(\mathbb{R}) : \text{sym}(S^\top A) = O \} = \{ A \in M_{n,p}(\mathbb{R}) : S^\top A + A^\top S = O \}, \]

\[ N_S\text{St}(p, n) := \{ SV : V \text{ is a } p \times p \text{ symmetric matrix} \}, \]

and the projection of \( X \) onto \( N_S\text{St}(p, n) \) is given by \( \text{Ssym}(S^\top X) \). Thus, the projection by \( \Pi : M_{n,p}(\mathbb{R}) \times \text{St}(p, n) \to T_S\text{St}(p, n) \) is written as

\[ \Pi(X, S) = X - \text{Ssym}(S^\top X) = S\text{sym}(S^\top X) + (I_n - SS^\top )X. \]

For further details for the Stiefel manifold, we refer to [18].

2.2. A first-order consensus model on \( \text{St}(p, n) \). In this subsection, we review the first-order model proposed in [41, 42] and study its basic property. First, we state the positive invariance of \( \text{St}(p, n) \) for (1.2) which can be guaranteed from the gradient flow structure.

**Lemma 2.1** (Positive invariance of the Stiefel manifold). Let \( S \) be a global solution to (1.2) with the initial data \( S^{\text{in}} := (S_{1}^{\text{in}}, \cdots, S_{N}^{\text{in}}) \). Then, we have

\[ S_{i}^{\text{in}} \in \text{St}(p, n) \implies S_{i}(t) \in \text{St}(p, n), \quad t > 0. \]

Next, we consider the left-translation invariance property whose proof directly follows from straightforward calculations.

**Lemma 2.2** (Left-translation invariance). For all \( L \in O(n) \), system (1.2) is invariant under left-translation by an \( n \times n \) orthogonal matrix in the sense that a transformed variable \( V_{i} := LS_{i} \) satisfies

\[ \dot{V}_{i} = \Omega_{i}V_{i} + V_{i}\Xi_{i} + \sum_{k=1}^{N} a_{ik} \left( V_{k} - \frac{1}{2}(V_{i}^\top V_{k} + V_{k}^\top V_{i}) \right). \]
The model (1.2) on the Stiefel manifold in fact includes several first-order models on the Riemannian manifolds such as $S_{n-1}$ in [45], $SO(n)$ in [53] and $S^1$ in [33]:

(i) $\dot{R}_i = \Omega_i R_i + \sum_{j=1}^N a_{ij} R_i \text{skew}(R_j^T R_j), \quad R_i \in SO(n).$

(ii) $\dot{x}_i = \Omega_i x_i + (I_n - x_i \otimes x_i) \sum_{j=1}^N a_{ij} x_j, \quad x_i \in S^{n-1}.$

(iii) $\dot{\theta}_i = \nu_i + \sum_{j=1}^N a_{ij} \sin(\theta_j - \theta_i), \quad \theta \in \mathbb{R}.$

Reduction basically follows from the property (2.1) and the projection operator, since the models (1.2) and (2.2) share the gradient flow structure. Moreover, system (1.2) satisfies the following splitting property for a homogeneous ensemble $\Xi_i \equiv \Xi$ for $i = 1, \ldots, N$. Then, the rotated variable $Y_i := S_i e^{-t\Xi}$ satisfies (1.2) with $\Xi \equiv O$. For a proof on the reduction and splitting property, we refer to Proposition 1 of [42]. Thus, when we consider a homogeneous ensemble, we set $\Xi \equiv O$ without loss of generality.

Remark 2.1. If we perform left-multiplication by $S_i^T$ in (1.2), then we obtain the following reduced dynamics:

$$S_i^T \dot{S}_i = \Xi_i + \frac{\kappa}{2N} \sum_{k=1}^N \left( S_i^T S_k - S_k^T S_i \right),$$

(2.3)

which means that (2.3) can be uniquely derived from (1.2). In other words, if $\{S_i\}_{i=1}^N$ is a solution to (1.2), then it also becomes a solution to (2.3). However, the converse statement might not hold, since there does not exist an inverse matrix of $S_i^T$. Hence, a solution set for (1.2) is a proper subset of that for (2.3).

Below, we briefly recall previous results in [41, 42] on the first-order model on the Stiefel manifolds. As far as the authors know, there has been only two literatures which concern with the consensus model (1.3) on the Stiefel manifolds. A graph $G$ is a pair $(V, E)$ where $V = \{1, \ldots, N\}$ and $E$ is a subset of $V$ where each subset is of cardinality two. Suppose that $S = (S_1, \ldots, S_N)$ is the state of particles interacting through the graph $G$.

Theorem 2.1. [41, 42] Suppose the pair $(p, n)$ satisfy

$$p \leq \frac{2}{3} n - 1,$$

and $G$ is connected. Let $S$ be a global solution to (1.1). Then, the consensus manifold $C$ defined by

$$C := \{(S_i)_{i=1}^N \in \text{St}(p, n)^N : S_i = S_j, \quad \forall \{i, j\} \in E\}$$

is almost globally asymptotically stable.
2.3. A second-order extension. In this subsection, we introduce a second-order extension (2.4) of (2.3) by adding suitable inertia like terms:

\[
m \ddot{S}_i = -mS_i \dot{S}_i^T \dot{S}_i - \gamma \dot{S}_i + S_i \Xi_i + \frac{m}{\gamma} \left(2 \ddot{S}_i \Xi_i - S_i \Xi_i S_i^T \dot{S}_i + S_i \dot{S}_i^T S_i \Xi_i \right)
\]

\[
+ \frac{\kappa}{N} \sum_{k=1}^{N} \left( S_k - \frac{1}{2} (S_i S_i^T S_k + S_i S_k^T S_i) \right).
\]

(2.4)

Note that in a formal zero inertia limit \( m \to 0 \), system (2.4) reduces to the first-order model (1.2). In order to show that the proposed model (2.4) is a suitable extension on the Stiefel manifold, we need to make sure that the governing manifold \( St(p, n) \) is still positively invariant along (2.4).

**Lemma 2.3** (Positive invariance of the Stiefel manifold). Suppose the initial data \((S_i^{in}, \dot{S}_i^{in})\) satisfy

\[
S_i^{in} \in St(p, n), \quad \dot{S}_i^{in} S_i^{in} + S_i^{in} \dot{S}_i^{in} = O, \quad i = 1, \ldots, N,
\]

and let \( S \) be a global solution to (2.4). Then, we have

\[
S_i(t) \in St(p, n), \quad i = 1, \ldots, N, \quad t \geq 0.
\]

**Proof.** We take the left multiplication of \( S_i^T \) to (2.4) to obtain

\[
m S_i^T \dot{S}_i = -m S_i^T S_i \dot{S}_i^T \dot{S}_i - \gamma S_i \dot{S}_i + S_i \Xi_i S_i^T \dot{S}_i + \frac{m}{\gamma} \left(2 S_i^T \ddot{S}_i \Xi_i - S_i^T S_i \Xi_i S_i^T \dot{S}_i + S_i^T \dot{S}_i S_i^T \Xi_i S_i \Xi_i \right)
\]

\[
+ \frac{\kappa}{2N} \sum_{k=1}^{N} \left[ S_i^T S_k - \frac{1}{2} (S_i^T S_i S_k^T S_i + S_i^T S_k S_i^T S_i) \right].
\]

(2.6)

We transpose (2.6) and use the skew-symmetry of \( \Xi_i \) to find

\[
m S_i^T \dot{S}_i = -m S_i^T \dot{S}_i^T S_i - \gamma S_i \dot{S}_i + S_i \Xi_i S_i^T \dot{S}_i + \frac{m}{\gamma} \left(-2 \Xi_i \dot{S}_i^T S_i + S_i^T S_i \Xi_i S_i^T \dot{S}_i - \Xi_i S_i^T \dot{S}_i S_i^T S_i \right)
\]

\[
+ \frac{\kappa}{2N} \sum_{k=1}^{N} \left[ S_k^T S_i - \frac{1}{2} (S_k^T S_i S_i^T S_k + S_k^T S_k S_i^T S_i) \right].
\]

(2.7)

We add (2.6) and (2.7) to get

\[
m (S_i^T \ddot{S}_i + \ddot{S}_i^T S_i) = -m (S_i^T \dot{S}_i \dot{S}_i^T \dot{S}_i + \dot{S}_i^T S_i S_i^T \dot{S}_i) - \gamma (S_i^T \ddot{S}_i + \ddot{S}_i^T S_i) + \left[ S_i^T S_i, \Xi_i \right]
\]

\[
+ \frac{m}{\gamma} J_1 + \frac{\kappa}{2N} \sum_{k=1}^{N} J_{2k},
\]

(2.8)

where \( J_1 \) and \( J_{2k} \) are defined as

\[
J_1 := 2 S_i^T \dot{S}_i \Xi_i - S_i^T S_i \Xi_i S_i^T \dot{S}_i + S_i^T \dot{S}_i S_i^T \dot{S}_i - 2 \Xi_i \dot{S}_i^T S_i + \dot{S}_i^T S_i \Xi_i S_i^T S_i - \Xi_i S_i^T \dot{S}_i S_i^T S_i,
\]

\[
J_{2k} := S_k^T S_k + S_k^T S_i - \frac{1}{2} \left( S_k^T S_i S_k^T S_i + S_k^T S_k S_i^T S_i + S_k^T S_k S_i S_i^T S_i + S_k^T S_k S_i S_i^T S_i \right).
\]

We recall the notation:

\[
H_i = I_p - S_i^T S_i, \quad \dot{H}_i = - (\dot{S}_i^T S_i + S_i^T \dot{S}_i) \quad \text{and} \quad \ddot{H}_i = - (\ddot{S}_i^T S_i + 2 \dot{S}_i^T \dot{S}_i + S_i^T \ddot{S}_i).
\]
In (2.9), we use the calculation of $J$ valued equation for $H_{\ddot{m}}$.

In what follows, we present the estimates for $J_1$ and $J_2$, respectively.

- (Estimate on $J_1$): By direct calculation, one has

$$
J_1 = 2\xi_i S_i^T \xi_i - \xi_i S_i^T \dot{S}_i + \dot{S}_i S_i \xi_i + H_i (\dot{S}_i^T S_i \xi_i - \xi_i S_i^T \dot{S}_i) - 2\xi_i \dot{S}_i S_i \xi_i - \xi_i S_i^T \dot{S}_i - (\dot{S}_i^T S_i \xi_i - \xi_i S_i^T \dot{S}_i) H_i - 2\dot{\xi}_i \dot{S}_i S_i \xi_i + H_i (\dot{S}_i^T S_i \xi_i - \xi_i S_i^T \dot{S}_i) - (\dot{S}_i^T S_i \xi_i - \xi_i S_i^T \dot{S}_i) H_i.
$$

- (Estimate on $J_{2k}$): Since the communication term of first-order and second-order models are same, it follows from Lemma 2.4 that

$$
J_{2k} = H_i (S_i^T S_k + S_k^T S_i) + (S_i^T S_k + S_k^T S_i) H_i.
$$

In (2.9), we use the calculation of $J_1$ and $J_2$ to find the second-order (autonomous) matrix-valued equation for $H_i$:

$$
m\dot{H}_i + \gamma \dot{H}_i + m H_i (\dot{S}_i^T \dot{S}_i) + m (\dot{S}_i^T \dot{S}_i) H_i - [H_i, \xi_i] + \frac{m}{\gamma} \left(2[\dot{\xi}_i, \dot{H}_i] + [H_i, \dot{S}_i^T S_i \xi_i - \xi_i S_i^T \dot{S}_i]\right) + \frac{K}{2N} \sum_{k=1}^{N} \left[H_i (S_i^T S_k + S_k^T S_i) + (S_i^T S_k + S_k^T S_i) H_i \right] = O.
$$

One can check that $H_i = O$ becomes a solution to (2.10) satisfying the initial assumption (2.5). Since a solution to the Cauchy problem (2.10) with the initial assumption (2.5) is unique, we conclude that

$$
H_i(t) = O, \quad t > 0.
$$

This yields the desired result.

Next, we focus on the situation in which solution operators for (2.4) can be expressed as a composition of two operators. In the proof of Lemma 2.3, we note that positive invariance of St$(p, n)$ is also valid, when the fourth term in the right-hand side of (2.4) is absent. However, the second-order model (2.4) satisfies such property for a homogeneous ensemble when the fourth term in the right-hand side of (2.4) is included.

Lemma 2.4. Suppose that the initial data and frequency matrices satisfy

$$
S_{i,1}^{\text{in}} \in \text{St}(p, n), \quad S_{i,1}^{\text{in}} \dot{S}_{i,1}^{\text{in}} + S_{i,1}^{\text{in}} \dot{S}_{i,1}^{\text{in}} = O, \quad \Xi_i \equiv \Xi, \quad i = 1, \cdots, N,
$$

and let $S$ be a global solution to (2.4). Then, $Y_i := S_i e^{-\frac{\Xi_i}{2}}$ satisfies

$$
m\dot{Y}_i = -m Y_i \dot{Y}_i - \gamma \dot{Y}_i + \frac{K}{N} \sum_{k=1}^{N} \left(Y_k - \frac{1}{2} (Y_k Y_k^T Y_k + Y_k^T Y_k) \right).
$$
Proof. By direct calculation, we use the ansatz for $Y_i$ to find its derivatives:

$$
\dot{Y}_i = \left( \dot{S}_i - \frac{S_i \Xi_i}{\gamma} \right) e^{-\frac{\Xi_i}{\gamma} t}, \quad \ddot{Y}_i = \left( \dot{S}_i - \frac{2\dot{S}_i \Xi_i}{\gamma} + \frac{S_i \Xi_i^2}{\gamma^2} \right) e^{-\frac{\Xi_i}{\gamma} t},
$$

$$
Y_i^T = e^{\frac{\Xi_i}{\gamma} t} S_i^T, \quad \dot{Y}_i^T = e^{\frac{\Xi_i}{\gamma} t} \left( \dot{S}_i^T + \frac{\Xi_i S_i^T}{\gamma} \right).
$$

In what follows, we consider the three terms:

$$m\ddot{Y}_i + mY_i \dot{Y}_i^T \dot{Y}_i, \quad \gamma \dot{Y}_i, \quad Y_k = \frac{1}{2} (Y_i Y_i^T Y_k + Y_i^T Y_i Y_k).$$

- (Estimate of $m\ddot{Y}_i + mY_i \dot{Y}_i^T \dot{Y}_i$): we observe

$$
m\ddot{Y}_i + mY_i \dot{Y}_i^T \dot{Y}_i = m \left( \ddot{S}_i - \frac{2\dot{S}_i \Xi_i}{\gamma} + \frac{S_i \Xi_i^2}{\gamma^2} \right) e^{-\frac{\Xi_i}{\gamma} t} + \frac{m}{\gamma} S_i \left( \dot{S}_i^T \dot{S}_i + \frac{\Xi_i S_i^T S_i - \Xi_i \Xi_i^T}{\gamma} \right) e^{-\frac{\Xi_i}{\gamma} t}.
$$

- (Estimate of $\gamma \dot{Y}_i$): By direct calculation,

$$
\gamma \dot{Y}_i = (\gamma \dot{S}_i - S_i \Xi_i) e^{-\frac{\Xi_i}{\gamma} t}.
$$

- (Estimate of $Y_k = \frac{1}{2} (Y_i Y_i^T Y_k + Y_i^T Y_i Y_k)$): we use the ansatz of $Y_i$ to find

$$
Y_k = \frac{1}{2} (Y_i Y_i^T Y_k + Y_i^T Y_i Y_k) = \left( S_k - \frac{1}{2} (S_i S_i^T S_k + S_k S_i^T S_i) \right) e^{-\frac{\Xi_i}{\gamma} t}.
$$

Finally, we combine the calculations above to find our desired result:

$$
m\ddot{Y}_i + mY_i \dot{Y}_i^T \dot{Y}_i + \gamma \dot{Y}_i - \frac{\kappa}{N} \sum_{k=1}^{N} \left( Y_k - \frac{1}{2} (Y_i Y_i^T Y_k + Y_i^T Y_i Y_k) \right)
$$

$$
= \left( m\ddot{S}_i + mS_i \dot{S}_i^T \dot{S}_i + \gamma \dot{S}_i - S_i \Xi_i - \frac{m}{\gamma} (2\dot{S}_i \Xi_i - S_i \Xi_i S_i^T \dot{S}_i + S_i \dot{S}_i^T S_i \Xi_i)
$$

$$
- \frac{\kappa}{N} \sum_{k=1}^{N} \left( S_k - \frac{1}{2} (S_i S_i^T S_k + S_k S_i^T S_i) \right) \right) e^{-\frac{\Xi_i}{\gamma} t} = O.
$$

We finally close this section by introducing second-order Grönwall-type inequalities.

Lemma 2.5. [9, 24] Let $y = y(t)$ be a nonnegative $C^2$-function satisfying the following differential inequality:

$$
a_i y + b_i y + c y \leq \varepsilon(t), \quad t > 0.
$$

Then, the following estimates hold:
(1) Suppose that $b^2 - 4ac > 0$ and $\varepsilon(t) \equiv \varepsilon_0$ is a given positive constant. Then, we have

\[
y(t) \leq \frac{\varepsilon_0}{c} + \left( y(0) + \frac{d}{c} \right) e^{-\nu_1 t} + \frac{a}{\sqrt{b^2 - 4ac}} \left( y'(0) + \nu_1 y(0) - \frac{2\varepsilon_0}{b - \sqrt{b^2 - 4ac}} \right) \left( e^{-\nu_2 t} - e^{-\nu_1 t} \right),
\]

where $\nu_1$ and $\nu_2$ are given as follows:

\[
\nu_1 := \frac{b + \sqrt{b^2 - 4ac}}{2a}, \quad \nu_2 := \frac{b - \sqrt{b^2 - 4ac}}{2a}.
\]

Then, one has

\[
\limsup_{t \to \infty} y(t) \leq \frac{\varepsilon_0}{c}.
\]

(2) Suppose that $b^2 - 4ac < 0$. Then, we have

\[
y(t) \leq \frac{4a\varepsilon_0}{b^2} + \left( y(0) - \frac{4a\varepsilon_0}{b^2} + \frac{b}{2a} y(0) + y'(0) - \frac{2\varepsilon_0}{b} \right) t \right) e^{-\frac{b}{2a} t}.
\]

Moreover,

\[
\limsup_{t \to \infty} y(t) \leq \frac{4a\varepsilon_0}{b^2}.
\]

(3) We furthermore assume that $\varepsilon(t)$ is a nonnegative continuously differentiable function decaying to zero as $t \to \infty$. Then, we have

\[
\lim_{t \to \infty} y(t) = 0.
\]

**Proof.** For the first two assertions, we refer the reader to [9, Lemma 3.1]. On the other hand for the last assertion, we refer the reader to [24, Lemma 4.9].

\[\square\]

3. First-order consensus model

In this section, we study the asymptotic behavior of the first-order model. In Section 3.1, we consider the homogeneous ensemble in which all natural frequencies are the same so that the complete consensus can be achieved. In Section 3.2, we are concerned with the heterogeneous ensemble where that phase-locked states can arise under some framework with a large coupling strength regime.

Recall the synchronization quantity $H_{ij} \in \text{M}_{p,p}(\mathbb{R})$:

\[
H_{ij} := I_p - S_i^\top S_j, \quad i, j = 1, \ldots, N.
\]

Then, we observe from the definition of $\text{St}(p,n)$

\[
\|S_i - S_j\|_F^2 = \text{tr}(I_p - S_i^\top S_j + I_p - S_j^\top S_i) = \text{tr}(H_{ij} + H_{ji}) \leq \sqrt{p} \|H_{ij} + H_{ji}\|_F, \quad (3.1)
\]

where we used the inequality:

\[
\text{tr}(AB) \leq \|A\|_F \|B\|_F \quad \text{for } A = I_p \text{ and } B = H_{ij} + H_{ji}.
\]

On the other hand, we see

\[
\|H_{ij} + H_{ji}\|_F = \|I_p - S_i^\top S_j + I_p - S_j^\top S_i\|_F = \|S_i^\top S_i - S_i^\top S_j + S_j^\top S_j - S_j^\top S_i\|_F \leq \|(S_i^\top - S_j^\top)(S_i - S_j)\|_F \leq \|S_i - S_j\|_F^2. \quad (3.2)
\]
In (3.1) and (3.2), one has
\[ \frac{1}{\sqrt{p}} \|S_i - S_j\|_F^2 \leq \|H_{ij} + H_{ji}\|_F \leq \|S_i - S_j\|_F^2. \] (3.3)

Thus, the following equivalence holds:
\[ \lim_{t \to \infty} \|S_i - S_j\|_F = 0 \iff \lim_{t \to \infty} \|H_{ij} + H_{ji}\|_F = 0. \]

3.1. A homogeneous ensemble. In this subsection, we consider a homogeneous ensemble where all frequency matrices are same:
\[ \Xi_i \equiv O, \quad i = 1, \ldots, N. \]

Lemma 3.1. Let \( S \) be a solution to (1.3) with \( \Xi_i \equiv \Xi \). Then, \( \|S_i - S_j\|_F^2 \) satisfies
\[
\frac{d}{dt} \|S_i - S_j\|_F^2 \leq -\frac{\kappa}{N} \sum_{k=1}^{N} a_{ik} (\|S_i - S_j\|^2 - \|S_j - S_k\|^2) - \frac{\kappa}{N} \sum_{k=1}^{N} a_{jk} (\|S_i - S_j\|^2 - \|S_i - S_k\|^2)
- (2 - \|S_i - S_j\|^2) \cdot \frac{\kappa}{2N} \sum_{k=1}^{N} (a_{ik} \|S_i - S_k\|^2 + a_{jk} \|S_j - S_k\|^2). \] (3.4)

Moreover, for the lower bound estimate, one has
\[ \frac{d}{dt} t^2_{ij} \geq -4\kappa a_{ij} t^2, \quad t > 0. \]

Proof. First, we derive the dynamics for \( \frac{d}{dt} (S_j^\top S_i) = \dot{S}_j^\top S_i + S_j^\top \dot{S}_i \). For this, we observe
\[
\dot{S}_j^\top S_i = S_j^\top S_i \Xi_i + \frac{\kappa}{N} \sum_{k=1}^{N} a_{ik} \left[ S_j^\top S_k - \frac{1}{2} (S_j^\top S_i S_k^\top S_i + S_j^\top S_k S_i^\top S_i) \right],
\]
\[
\dot{\Xi}_j S_i = -\Xi_j S_j^\top S_i + \frac{\kappa}{N} \sum_{k=1}^{N} a_{jk} \left[ S_k^\top S_i - \frac{1}{2} (S_k^\top S_j S_i^\top S_j + S_k^\top S_i S_j^\top S_i) \right]. \] (3.5)

Then, we calculate (3.5)_1 + (3.5)_2 to find the dynamics for \( A_{ji} := S_j^\top S_i \):
\[
\frac{d}{dt} A_{ji} = A_{ji} \Xi_i - \Xi_j A_{ji}
+ \frac{\kappa}{N} \sum_{k=1}^{N} a_{ik} \left[ A_{jk} - \frac{1}{2} (A_{ji} A_{ik} + A_{ki} A_{ji}) \right] + a_{jk} \left[ A_{ki} - \frac{1}{2} (A_{kj} A_{ji} + A_{jk} A_{ji}) \right]. \] (3.6)

Again, if we recall the notation:
\[ H_{ji} = I_p - A_{ji} = I_p - S_j^\top S_i, \]
then (3.6) directly yields the dynamics for $H_{ji}$:

$$
\frac{d}{dt}H_{ji} = (\Xi_j - \Xi_i) + (H_{ji}\Xi_i - \Xi_jH_{ji}) - \frac{\kappa}{N} \sum_{k=1}^{N} (a_{ik} + a_{jk})H_{ji}
$$

$$
+ \frac{\kappa}{2N} \sum_{k=1}^{N} a_{ik}(2H_{jk} - H_{ik} - H_{kj}) + a_{ik}(H_{ji}H_{ik} + H_{ji}H_{ki})
$$

$$
+ \frac{\kappa}{2N} \sum_{k=1}^{N} a_{jk}(2H_{ki} - H_{jk} - H_{kj}) + a_{jk}(H_{kj}H_{ji} + H_{jk}H_{ji}).
$$

(3.7)

We interchange the index $i \leftrightarrow j$ in (3.7) to get

$$
\frac{d}{dt}H_{ij} = (\Xi_i - \Xi_j) + (H_{ij}\Xi_j - \Xi_iH_{ij}) - \frac{\kappa}{N} \sum_{k=1}^{N} (a_{jk} + a_{ik})H_{ij}
$$

$$
+ \frac{\kappa}{2N} \sum_{k=1}^{N} a_{jk}(2H_{ik} - H_{jk} - H_{kj}) + a_{jk}(H_{ij}H_{jk} + H_{ij}H_{kj})
$$

$$
+ \frac{\kappa}{2N} \sum_{k=1}^{N} a_{ik}(2H_{kj} - H_{ik} - H_{ki}) + a_{ik}(H_{ki}H_{ij} + H_{ki}H_{ij}).
$$

(3.8)

We add (3.7) and (3.8) to find

$$
\frac{d}{dt}(H_{ij} + H_{ji}) = -(\Xi_i - \Xi_j) + (H_{ij}\Xi_i - \Xi_jH_{ji} + H_{ji}\Xi_j - \Xi_iH_{ij})
$$

$$
+ \frac{\kappa}{2N} \sum_{k=1}^{N} (a_{ik} + a_{jk})(H_{ij} + H_{ji}) + (H_{ij}\Xi_i - \Xi_jH_{ji} + H_{ji}\Xi_j - \Xi_iH_{ij})
$$

$$
+ \frac{\kappa}{2N} \sum_{k=1}^{N} a_{jk}(2H_{ik} - H_{jk} - H_{kj}) + a_{jk}(H_{ij}H_{jk} + H_{ij}H_{kj})
$$

$$
+ \frac{\kappa}{2N} \sum_{k=1}^{N} a_{ik}(2H_{kj} - H_{ik} - H_{ki}) + a_{ik}(H_{ki}H_{ij} + H_{ki}H_{ij})
$$

$$
+ a_{jk}((H_{jk} + H_{kj})H_{ji} + H_{ij}(H_{jk} + H_{kj})).
$$

(3.9)

We take the trace in (3.9) to find

$$
\frac{d}{dt}\|S_i - S_j\|^2 = -\frac{\kappa}{N} \sum_{k=1}^{N} (a_{ik} + a_{jk})\|S_i - S_j\|^2 + \text{tr}((H_{ij} - H_{ji})(\Xi_i - \Xi_j))
$$

$$
+ \frac{\kappa}{2N} \sum_{k=1}^{N} (a_{jk} - a_{ik})(\|S_i - S_k\|^2 - \|S_j - S_k\|^2)
$$

$$
+ \frac{\kappa}{2N} \sum_{k=1}^{N} \text{tr}[(a_{ik}(H_{ik} + H_{ki}) + a_{jk}(H_{jk} + H_{kj}))(H_{ij} + H_{ji})].
$$

(3.10)
Finally, we use $\Xi_i \equiv O$ and the inequality (3.3) to estimate the last term in (3.10):

$$\frac{d}{dt} \|S_i - S_j\|^2 \leq -\frac{\kappa}{N} \sum_{k=1}^N a_{ik}(\|S_i - S_j\|^2 - \|S_j - S_k\|^2) - \frac{\kappa}{N} \sum_{k=1}^N a_{jk}(\|S_i - S_j\|^2 - \|S_i - S_k\|^2)$$

$$- (2 - \|S_i - S_j\|^2) \cdot \frac{\kappa}{2N} \sum_{k=1}^N (a_{ik}\|S_i - S_k\|^2 + a_{jk}\|S_j - S_k\|^2).$$

\[\Box\]

We are now ready to prove a proof of Theorem 1.1. For this, we define the maximal diameter:

$$D(S(t)) := \max_{1 \leq i,j \leq N} \|S_i(t) - S_j(t)\|_F, \quad t \geq 0.$$

(Proof of Theorem 1.1): For each $t > 0$, we choose the maximal index $(i_t, j_t)$:

$$D(S(t)) = \|S_{i_t} - S_{j_t}\|_F.$$

Then, it follows from (3.4) in Lemma 3.1 that $D(S)$ satisfies

$$\frac{d}{dt} D(S)^2 \leq -(2 - D(S)^2) \cdot \frac{\kappa}{2N} \sum_{k=1}^N (a_{ik}d_{i,k}^2 + a_{jk}d_{j,k}^2) \leq -\frac{\kappa a_m}{4}(2 - D(S)^2)D(S)^2.$$

Thus, the region $\{t \geq 0 : D(S(t)) < \sqrt{2}\}$ is positively invariant and the desired assertion directly follows from the dynamical systems theory. \[\Box\]

**Remark 3.1.** In Theorem 2.1, their initial framework leading to the complete consensus only depends on the relation of $(p, n)$ (in fact, $p \leq \frac{2n}{3} - 1$), and the initial data $S^{in}$ is not restricted except for a measure zero set, so that the global consensus is achieved. In contrast, for our result, any $(p, n)$ can be chosen; however, as a trade-off, smallness on the initial data would be imposed, for instance, $D(S^{in}) < \sqrt{2}$.  

### 3.2. A heterogeneous ensemble.

In this subsection, we deal with the heterogeneous ensemble, and provide a sufficient framework leading to the phase-locked state. In order to capture the phase-locked state, we briefly present our strategy introduced in [28] consisting of several steps:

- **Step A:** let $S$ and $\tilde{S}$ two solutions to (1.3). Derive a temporal evolution of the maximal quantity (inter-diameter) denoted by $d(S, \tilde{S})$:

$$d(S, \tilde{S}) := \max_{1 \leq i,j \leq N} \|S_i^\top S_j - \tilde{S}_i^\top \tilde{S}_j\|_F = \max_{1 \leq i,j \leq N} \|A_{ij} - \tilde{A}_{ij}\|_F,$$

which measures the relative distance between two solution configurations $S$ and $\tilde{S}$ (see Lemma 3.2).

- **Step B:** in Step A, since we need a smallness on the diameter $D(S)$, we find a positively invariant region of $D(S)$ (see Lemma 3.3).

- **Step C:** together with a temporal evolution of $d(S, \tilde{S})$ and a positively invariant region of $D(S)$, we show that $d(S, \tilde{S})$ converges to zero. Then, since our system is autonomous, $\tilde{S}_i(t) = S_i(t + T)$ also becomes a solution to the system for any $T$. By discretizing the time $t \in \mathbb{R}_+$ as $n \in \mathbb{Z}_+$ and setting $T = m \in \mathbb{Z}_+$, we deduce that $\{S_j^\top (n)S_i(n)\}_{n \in \mathbb{Z}_+}$ is indeed a
Cauchy sequence in the radius $p$-ball of $M_p(\mathbb{R})$ and hence each $S_j^T S_i$ converges to a constant $p \times p$ matrix.

In what follows, we provide the detailed justification of Step A, Step B and Step C. For this, we assume

$$\Lambda := a_m - \frac{N - 1}{N}(a_M + d(A)) \in (0, 8pa_M^2).$$

In other words, difference between the maximum and minimum would be small. Below, we study the temporal evolution of $d(S, \tilde{S})$ in the following lemma.

**Lemma 3.2.** Let $S$ and $\tilde{S}$ be two solutions to (1.3). Then, the inter-diameter $d(S, \tilde{S})$ satisfies

$$\frac{d}{dt} d(S, \tilde{S}) \leq -2\kappa (\Lambda - a_M \sqrt{p}(D(S) + D(\tilde{S})))d(S, \tilde{S}), \quad t > 0.$$

**Proof.** Since the proof is lengthy, we leave its proof in Appendix A. □

Next, we find a positively invariant region of $D(S)$. For this, we define $0 < \alpha < \beta$ through the relation:

$$\alpha \text{ and } \beta \text{ are two positive roots of } r^3 - 2r + \frac{2\sqrt{p}\|\Xi\|_\infty}{\kappa a_m} = 0,$$

where $\kappa_*$ is a positive constant defined by

$$\kappa_* := \frac{16p^2a_M^3\|\Xi\|_\infty}{a_m(8pa_M^2\Lambda - \Lambda^3)}.$$

**Lemma 3.3.** Suppose that the coupling strength and the initial data satisfy

$$\kappa > \max \left\{ \frac{\sqrt{p}\|\Xi\|_\infty}{9a_m}, \frac{16p^2a_M^3\|\Xi\|_\infty}{a_m(8pa_M^2\Lambda - \Lambda^3)} \right\}, \quad D(S^{\text{in}}) < \beta, \quad (3.11)$$

and let $S$ a global solution to (1.3). Then, there exists a finite entrance time $T_2 > 0$ such that

$$D(S(t)) < \alpha < \frac{\Lambda}{2a_M \sqrt{p}}, \quad t > T_2.$$

**Proof.** Recall the estimate (3.4) in Lemma 3.1 and find the differential inequality of $D(S)$ for a heterogeneous ensemble:

$$\frac{d}{dt} D(S) \leq -\frac{\kappa a_m}{2}D(S) + \frac{\kappa a_m}{4}D(S)^3 + \frac{\sqrt{p}||\Xi||_\infty}{2}$$

$$= \frac{\kappa a_m}{4} \left( -2D(S) + D(S)^3 + \frac{2\sqrt{p}||\Xi||_\infty}{\kappa a_m} \right) =: f(D(S)),$$

where we introduce an auxiliary cubic function $f$:

$$f(r) = r^3 - 2r + \frac{2\sqrt{p}||\Xi||_\infty}{\kappa a_m}, \quad r \geq 0.$$

Then, it follows from the simple calculus that

$$f \text{ attains the minimum at } r_* := \sqrt{\frac{2}{3}} \text{ with } f(r_*) = \frac{2\sqrt{p}||\Xi||_\infty}{\kappa a_m} - \frac{2}{3} \sqrt{\frac{2}{3}}.$$
Thus, since $f(r_*) < 0$ is due to $\text{(3.11)}_1$, we see that $f$ has two positive roots $r_1$ and $r_2$ such that
\[ 0 < r_1 < \sqrt{\frac{2}{3}}, \quad \sqrt{\frac{2}{3}} < r_2 < \sqrt{2}. \]
Note that $\text{(3.11)}_1$ indeed gives $f(r_*) < 0$. In addition, we observe
\[ \lim_{t \to \infty} r_1 = 0 \quad \lim_{t \to \infty} r_2 = \sqrt{2}. \]
I use the dynamical systems theory to show that if $D(S^{\text{in}}) < r_2$, then there exists a finite entrance time $T_3 > 0$ such that
\[ D(S(t)) < r_1, \quad t > T_3. \]
More precisely, we split the case into two parts. First, suppose initial data satisfy
\[ D(S^{\text{in}}) \leq r_1. \]
Then, at the time $t = t_*$ when $D(S(t_*)) = r_1$, we have
\[ \frac{d}{dt}D(S) \leq 0. \]
Thus, $D(S)$ does not increase at $t = t_*$ and hence is restricted in the interval $[0, r_1]$. Second, suppose that the initial data satisfy $r_1 < D(S^{\text{in}}) < r_2$. Again, at the instant time $t = t_*$ when $r_1 < D(S(t_*)) < r_2$,
\[ \frac{d}{dt}D(S) < 0. \]
Thus, $D(S)$ starts to strictly decrease. Then, by applying same argument in Proposition 3.1 in [8], we find such finite entrance time $T_3 > 0$.

In order to find the desired invariant region, we have to assume
\[ f \left( \frac{\Lambda}{2aM\sqrt{p}} \right) = \left( \frac{\Lambda}{2aM\sqrt{p}} \right)^3 - 2 \cdot \left( \frac{\Lambda}{2aM\sqrt{p}} \right) + \frac{2\sqrt{p}\|\Xi\|_{\infty}}{\kappa a_m} < 0, \quad \text{(3.12)} \]
and one can check from algebraic manipulation that the relation $\text{(3.12)}$ is equivalent to $\text{(3.11)}_1$. Then, the cubic equation
\[ r^3 - 2r + \frac{2\sqrt{p}\|\Xi\|_{\infty}}{\kappa a_m} = 0 \]
has two positive roots: $\alpha$ and $\beta$. Hence, under the assumption $\text{(3.11)}$, there exists such finite entrance time $T_2 > 0$ so that
\[ D(S(t)) < \alpha < \frac{\Lambda}{2aM\sqrt{p}}, \quad t > T_2. \]
Finally, we use Lemma 3.2 and Lemma 3.3 to present a proof of Theorem 1.2.

(Proof of Theorem 1.2): Since we assume $\text{(3.11)}$, we use Lemmas 3.2 and 3.3 to show that
\[ \frac{d}{dt}d(S, \tilde{S}) \leq -2\kappa(\Lambda - 2aM\sqrt{p}\alpha)d(S, \tilde{S}), \quad \text{a.e. } t > T_2. \]
By induction argument, (3.15) gives for
(3.13) yields
Thus, we find
For two solutions
Proof. Especially for \( T = 1 \) and \( t = n \in \mathbb{Z}_+ \) in (3.14), we have
By induction argument, (3.15) gives for \( m \in \mathbb{Z}_+ \),
Hence, the discretized sequence \( \{ S_j^T(n)S_i(n) \}_{n \in \mathbb{Z}_+} \) is indeed Cauchy in the \( p \)-ball \( B_p(O) := \{ X \in M_p(\mathbb{R}) : \| X \|_F \leq p \} \). Consequently, for each \( i,j \), it converges to a constant \( p \times p \) matrix \( \Gamma_{ji}^{\infty} \in B_p(O) \).
As a corollary, we show that normalized velocities synchronize.

**Corollary 3.1.** Suppose that the coupling strength and initial data satisfy (3.11) and let \( S \) be a global solution to (1.3). Then, the normalized velocities \( S_i^T \dot{S}_i \) and \( S_j^T \dot{S}_j \) synchronize:

\[
\lim_{t \to \infty} \| S_i^T \dot{S}_i - S_j^T \dot{S}_j \|_F = 0.
\]

**Proof.** For two solutions \( S \) and \( \tilde{S} \), we observe from (2.3)

\[
S_i^T \dot{S}_i - S_i^T \tilde{S}_i = \Xi_i + \frac{\kappa}{2N} \sum_{k=1}^{N} a_{ik} (S_i^T S_k - \tilde{S}_i^T S_k) S_i - \Xi_i + \frac{\kappa}{2N} \sum_{k=1}^{N} a_{ik} (\tilde{S}_i^T \tilde{S}_k - \tilde{S}_i^T \tilde{S}_k) - \Xi_i + \frac{\kappa}{2N} \sum_{k=1}^{N} a_{ik} (\tilde{S}_i^T \tilde{S}_k - \tilde{S}_i^T \tilde{S}_k) - \Xi_i + \frac{\kappa}{2N} \sum_{k=1}^{N} a_{ik} (\tilde{S}_i^T \tilde{S}_k - \tilde{S}_i^T \tilde{S}_k) - \Xi_i + \frac{\kappa}{2N} \sum_{k=1}^{N} a_{ik} (\tilde{S}_i^T \tilde{S}_k - \tilde{S}_i^T \tilde{S}_k)
\]

Thus, we find

\[
\| S_i^T \dot{S}_i - S_i^T \tilde{S}_i \|_F \leq \frac{\kappa}{2N} \sum_{k=1}^{N} a_{ik} (\| S_i^T S_k - \tilde{S}_i^T S_k \| + \| S_i^T S_k - \tilde{S}_i^T S_k \|) \leq \kappa a_M d(S, \tilde{S}).
\]

Finally, we use the same argument in Theorem 1.2 to show that \( \{ S_i^T \dot{S}_i \} \) becomes a Cauchy sequence and hence it converges to the same constant matrix. This shows the desired result. \( \square \)
4. A second-order consensus model

In this section, we study the emergent dynamics of the second-order model (1.4). Before we present the estimates, energy functionals are defined and their temporal evolutions are derived as a crucial step. After that, the homogeneous ensemble are considered in Section 4.1 and we provide a sufficient framework leading to the complete consensus. On the other hand, we deal with the heterogeneous ensemble in Section 4.2 and present a sufficient framework for the practical consensus in the large coupling strength and small inertia regime.

As in [25], we define a total energy functional associated to (1.4):

$$
\mathcal{E} = \frac{m}{N} \sum_{i=1}^{N} \| \dot{S}_i \|_F^2 + \frac{\kappa}{2N^2} \sum_{i,j=1}^{N} a_{ij} \| S_i - S_j \|_F^2,
$$

where $K$ represents a (total) kinetic energy and $L$ describes the interaction energy between the states of particles $\{ S_i \}$ measuring the degree of consensus.

Next, we study the temporal-evolution of the total energy $\mathcal{E}$. First, we differentiate $\| \dot{S}_i \|_F^2$ with respect to $t$ to find

$$
m \frac{d}{dt} \| \dot{S}_i \|_F^2 = m \frac{d}{dt} \text{tr}( \dot{S}_i^T \dot{S}_i ) = m \text{tr}( \dot{S}_i^T \dot{S}_i + \dot{S}_i^T \dot{S}_i ) = m \text{tr}( \dot{S}_i^T \dot{S}_i + (\dot{S}_i^T \dot{S}_i)^T )
$$

First, we observe the term $m \dot{S}_i^T \dot{S}_i$:

$$
m \dot{S}_i^T \dot{S}_i = -m \dot{S}_i^T S_i^T \dot{S}_i - \gamma \dot{S}_i^T S_i - \dot{S}_i^T \dot{S}_i + \frac{m}{\gamma} (2 \dot{S}_i^T \dot{S}_i \Xi_i - \dot{S}_i^T \dot{S}_i \Xi_i + \dot{S}_i^T S_i \dot{S}_i S_i \dot{S}_i \Xi_i)
$$

$$
+ \frac{\kappa}{N} \sum_{k=1}^{N} a_{ik} \left( \dot{S}_i^T S_k - \frac{1}{2} (\dot{S}_i^T S_i \dot{S}_i + S_i^T S_i \dot{S}_i + S_i^T S_i \dot{S}_i) \right)
$$

$$
= m I_1 - \gamma I_2 + I_3 + \frac{m}{\gamma} I_4 + \frac{\kappa}{N} \sum_{k=1}^{N} I_5.
$$

(4.1)

In next lemma, we provide estimates for $I_k$, $k = 1, \ldots, 5$.

**Lemma 4.1.** Let $I_k$, $k = 1, \ldots, 5$ be the terms defined in (4.1). Then, one has

$$
\text{tr} \left( I_1 + I_1^T \right) = 0, \quad \text{tr} \left( I_2 + I_2^T \right) = 2 \| \dot{S}_i \|_F^2, \quad \text{tr} \left( I_3 + I_3^T \right) = \text{tr} \left( \dot{S}_i^T S_i \Xi_i - \Xi_i S_i^T S_i \dot{S}_i \right)
$$

$$
\text{tr} \left( I_4 + I_4^T \right) = 0, \quad \text{tr} \left( I_5 + I_5^T \right) = a_{ik} \text{tr} \left( S_i^T S_k + S_k^T S_i \right).
$$

**Proof.** Below, we present estimates of $\text{tr}(I_k + I_k^T)$, $k = 1, \ldots, 5$, respectively.

- **(Estimate of $\text{tr}(I_1 + I_1^T)$):** we recall the identity $\dot{S}_i^T S_i + S_i^T \dot{S}_i = O_p$ to see

$$
\text{tr} \left( I_1 + I_1^T \right) = \text{tr}(\dot{S}_i^T S_i \dot{S}_i + \dot{S}_i^T \dot{S}_i) = \text{tr} \left[ \dot{S}_i^T \dot{S}_i \right] = 0.
$$

- **(Estimate of $\text{tr}(I_2 + I_2^T)$):** by direct observation,

$$
\text{tr} \left( I_2 + I_2^T \right) = 2 \text{tr} (I_2) = 2 \| \dot{S}_i \|_F^2.
$$
(Estimate of \( I_3 + I_3^T \)): we use the property of the trace \( \text{tr}(AB) = \text{tr}(BA) \) to find
\[
\text{tr}(I_3 + I_3^T) = \text{tr}\left(\dot{S}_i^T S_i \Xi_i - \Xi_i S_i^T \dot{S}_i\right).
\]

(Estimate of \( I_4 + I_4^T \)): we first see
\[
\text{tr}(I_4) = \text{tr}\left((2\dot{S}_i^T \dot{S}_i - S_i^T \dot{S}_i \dot{S}_i^T S_i + \dot{S}_i^T S_i \dot{S}_i^T S_i)\Xi_i\right)
= \text{tr}\left((2\dot{S}_i^T \dot{S}_i - S_i^T \dot{S}_i \dot{S}_i^T S_i - \dot{S}_i^T S_i \dot{S}_i^T S_i)\Xi_i\right).
\]
(4.2)
Since the matrix \( 2\dot{S}_i^T \dot{S}_i - S_i^T \dot{S}_i \dot{S}_i^T S_i - \dot{S}_i^T S_i \dot{S}_i^T S_i \) in (4.2) is symmetric and \( \Xi_i \) is skew-symmetric, we obtain
\[
\text{tr}(I_4 + I_4^T) = 0.
\]

(Estimate of \( I_5 + I_5^T \)): we again use the identity \( \dot{S}_i^T S_i + S_i^T \dot{S}_i = 0 \) to see
\[
\text{tr}(I_5 + I_5^T)
= a_{ik}\text{tr}(\dot{S}_i^T S_k + S_k^T \dot{S}_i) - \frac{a_{ik}}{2}\text{tr}(\dot{S}_i^T S_i S_k^T S_k + \dot{S}_i^T S_i S_k^T \dot{S}_i + S_k^T S_i \dot{S}_i S_k^T S_k + S_k^T S_i \dot{S}_i S_k^T S_k)
= a_{ik}\text{tr}(\dot{S}_i^T S_k + S_k^T \dot{S}_i) - \frac{a_{ik}}{2}\text{tr}[(\dot{S}_i^T S_i + S_i^T \dot{S}_i)(S_k^T S_k + S_k^T S_k)]
= a_{ik}\text{tr}(\dot{S}_i^T S_k + S_k^T \dot{S}_i).
\]

Next, we provide an estimate on the time-rate of change of the total energy.

**Proposition 4.1.** Let \( S \) be a global solution to (1.4). Then, the total energy functional \( E \) satisfies
\[
\frac{dE}{dt} = -\frac{2\gamma}{N} \sum_{i=1}^{N} ||\dot{S}_i||_F^2 + \frac{1}{N} \sum_{i=1}^{N} \text{tr}(\dot{S}_i^T S_i \Xi_i - \Xi_i S_i^T \dot{S}_i).
\]
(4.3)

**Proof.** We collect all estimates of \( \text{tr}(I_i + I_i^T) \) for \( i = 1, \ldots, 5 \) to find
\[
m\frac{d}{dt}||\dot{S}_i||_F^2 = -2\gamma||\dot{S}_i||_F^2 + \text{tr}(\dot{S}_i^T S_i \Xi_i - \Xi_i S_i^T \dot{S}_i) + \frac{K}{N} \sum_{k=1}^{N} a_{ik}\text{tr}(\dot{S}_i^T S_k + S_k^T \dot{S}_i),
\]
(4.4)
and sum (4.4) over all \( i \) and divide the resulting relation by \( N \) to obtain
\[
m\frac{1}{N} \frac{d}{dt} \sum_{i=1}^{N} ||\dot{S}_i||_F^2 = -\frac{2\gamma}{N} \sum_{i=1}^{N} ||\dot{S}_i||_F^2 + \frac{1}{N} \sum_{i=1}^{N} \text{tr}(\dot{S}_i^T S_i \Xi_i - \Xi_i S_i^T \dot{S}_i) + K\text{tr}(\dot{S}_c^T S_c + S_c^T \dot{S}_c)
= -\frac{2\gamma}{N} \sum_{i=1}^{N} ||\dot{S}_i||_F^2 + \frac{1}{N} \sum_{i=1}^{N} \text{tr}(\dot{S}_i^T S_i \Xi_i - \Xi_i S_i^T \dot{S}_i) + \frac{K}{N^2} \sum_{i,k=1}^{N} a_{ik}\text{tr}(\dot{S}_i^T S_k + S_k^T \dot{S}_i).
\]
(4.5)
On the other hand, we observe
\[
\frac{\kappa}{N^2} \sum_{i,k=1}^{N} a_{ik} \text{tr}(\dot{S}_i^T S_k + S_k^T \dot{S}_i) = \frac{\kappa}{2N^2} \sum_{i,k=1}^{N} a_{ik} \text{tr}(\dot{S}_i^T S_k + S_k^T \dot{S}_i + S_i^T S_k + S_k^T S_i)
\]
\[
= \frac{\kappa}{2N^2} \frac{d}{dt} \left( \sum_{i,k=1}^{N} a_{ik} \text{tr}(S_i^T S_k + S_k^T S_i) \right) = -\frac{\kappa}{2N^2} \frac{d}{dt} \left( \sum_{k=1}^{N} a_{ik} \|S_i - S_k\|^2 \right).
\]
Finally in (4.5), we obtain the desired dissipation energy estimate. □

Remark 4.1. It follows from (4.4) that
\[
m \frac{d}{dt} \|\dot{S}_i\|_F^2 \leq -2\gamma \|\dot{S}_i\|_F^2 + 2\|\dot{S}_i\|_F \|S_i\|_F + 2\kappa a_M \sqrt{p} \|\dot{S}_i\|_F, \quad t > 0,
\]
or equivalently
\[
m \frac{d}{dt} \|\dot{S}_i\|_F \leq -\gamma \|\dot{S}_i\|_F + \|\Xi\|_F + \kappa a_M \sqrt{p}, \quad t > 0.
\]
We set the maximal value of \(\|\Xi\|_F\):
\[
\|\Xi\|_\infty := \max_{1 \leq i \leq N} \|\Xi_i\|_F.
\]
Then, Grönwall’s lemma yields a uniform boundedness and uniform continuity of \(\|\dot{S}_i\|_F\):
\[
\sup_{0 \leq t < \infty} \|\dot{S}_i(t)\|_F \leq \max \left\{ \max_{1 \leq i \leq N} \|\dot{S}_{i,n}\|_F, \frac{1}{\gamma} \left( \|\Xi\|_\infty + \kappa a_M \sqrt{p} \right) \right\}.
\]
(4.6)

In what follows, we impose the condition on the network topology:
\[
\frac{1}{N} \sum_{k=1}^{N} a_{ik} \equiv \xi > 0, \quad i = 1, \cdots, N.
\]
In other words, every columns (or rows) have a common average \(\xi\). We denote several quantities:
\[
\mathcal{G}(t) := \frac{1}{N^2} \sum_{i,j=1}^{N} \|S_i(t) - S_j(t)\|^2_F, \quad D(\dot{S}(t)) := \max_{1 \leq i \leq N} \|\dot{S}_i(t)\|_F, \quad \|\Xi\|_\infty := \max_{1 \leq i \leq N} \|\Xi_i\|_F.
\]
Below, we derive a second-order differential inequality for \(\mathcal{G}\).

Lemma 4.2. Let \(S\) be a solution to (1.4). Then, \(\mathcal{G}\) satisfies
\[
m \ddot{\mathcal{G}} + \gamma \dot{\mathcal{G}} + 2\kappa \xi \mathcal{G} \leq 16mD(\dot{S})^2 + 8\|\Xi\|_\infty + \frac{16m\sqrt{p} \|\Xi\|_\infty}{\gamma} D(\dot{S}).
\]
(4.7)
Proof. Since the proof is lengthy, we postpone the justification of (4.7) in Appendix B. □

Below, we quote Barbalat’s lemma without proofs.

Lemma 4.3. \(\Box\)

(i) Suppose that a real-valued function \(f : [0, \infty) \rightarrow \mathbb{R}\) is uniformly continuous and satisfies
\[
\lim_{t \to \infty} \frac{1}{t} \int_{0}^{t} f(s) \, ds \text{ exists.}
\]
Then, $f$ tends to zero as $t \to \infty$:

$$
\lim_{t \to \infty} f(t) = 0.
$$

(ii) Suppose that a real-valued function $f : [0, \infty) \to \mathbb{R}$ is continuously differentiable, and $\lim_{t \to \infty} f(t) = \alpha \in \mathbb{R}$. If $f'$ is uniformly continuous, then

$$
\lim_{t \to \infty} f'(t) = 0.
$$

4.1. A homogeneous ensemble. In this part, we consider a homogeneous ensemble whose frequency matrices are the same, i.e., $\Xi_i \equiv \Xi$. Due to the property in Lemma 2.4 for a homogeneous ensemble, without loss of generality, we may set

$$
\Xi_i \equiv O, \quad i = 1, \ldots, N.
$$

We are now ready to provide a proof of Theorem 1.3.

(Proof of Theorem 1.3) For the zero convergence of $\|\dot{S}_i\|_F$, we use in (4.3) the condition $\Xi_i \equiv O$ to get

$$
\frac{dE}{dt} = -\frac{2\gamma}{N} \sum_{i=1}^{N} \|\dot{S}_i\|_F^2, \quad t > 0.
$$

We integrate the above equation on the interval $[0, t]$ to obtain

$$
\frac{2\gamma}{N} \int_{0}^{t} \sum_{i=1}^{N} \|\dot{S}_i(s)\|_F^2 ds = E(0) - E(t) \leq E(0).
$$

Hence, one has

$$
\int_{0}^{\infty} \|\dot{S}_i(t)\|_F^2 dt < \infty.
$$

It follows from Remark 4.1 that $\left| \frac{d}{dt} \|\dot{S}_i\|_F \right|$ is uniformly bounded, which implies the uniform continuity of $\|\dot{S}_i\|_F$. Hence, we can apply Barbalat’s lemma [1] to derive the desired zero convergence of $\|\dot{S}_i\|_F$. On the other hand for the zero convergence of $G$, we recall the differential inequality (4.7). Then, since we have $\|\Xi\|_\infty = 0$ and $\mathcal{D}(\dot{S})$ converges to zero, we apply Lemma 2.5 to obtain the desired result. $\square$

4.2. A heterogeneous ensemble. In this part, we study a heterogeneous ensemble and present a proof of Theorem 1.4. For this, we impose a small inertia assumption such that there exist $m_0 > 0$ which does not depend on $\kappa$ and $\eta > 0$ satisfying

$$
m = \frac{m_0}{\kappa^{1+\eta}}.
$$

Since $\kappa$ will tend to infinity, the relation above is referred as a small inertial assumption. The reason why we need such assumption is clarified throughout the proof.

(Proof of Theorem 1.4): since we assume (4.7), it follows from (4.6) that

$$
\mathcal{D}(\dot{S}(t)) < \frac{1}{\gamma}(\|\Xi\|_\infty + \kappa a_M \sqrt{p}).
$$
Then, the right-hand side of (4.7) becomes
\[16m D(\dot{S})^2 + 8\|\Xi\|_\infty + \frac{16m\sqrt{p}\|\Xi\|_\infty}{\gamma} D(\dot{S})\]
\[\leq \frac{16m}{\gamma^2} (\|\Xi\|_\infty + \kappa a M \sqrt{p})^2 + \frac{16m\sqrt{p}\|\Xi\|_\infty}{\gamma^2} (\|\Xi\|_\infty + \kappa a M \sqrt{p} \|\Xi\|_\infty) + 8\|\Xi\|_\infty\]
\[= \frac{16m}{\gamma^2} (\|\Xi\|_\infty + \kappa a M \sqrt{p} (\|\Xi\|_\infty + \kappa a M \sqrt{p} + \sqrt{p}\|\Xi\|_\infty)) + 8\|\Xi\|_\infty\]
\[= O(m\kappa^2) + 8\|\Xi\|_\infty = O(m\kappa^2) + O(1).\]

In order to apply second-order Grönwall’s inequality in Lemma 2 is used, we consider two cases:

(i) \(\gamma^2 - 8\xi m\kappa > 0\).
(ii) \(\gamma^2 - 8\xi m\kappa < 0\).

For the first case, we have
\[\limsup_{t \to \infty} G(t) \leq \frac{1}{2\xi \kappa} (O(m\kappa^2) + O(1)) = O(m\kappa) + O(\kappa^{-1}) = O(\kappa^{-\eta}) + O(\kappa^{-1}),\]
where we used the relation (1.7) in the last equality. On the other hand for the second case, we have
\[\limsup_{t \to \infty} G(t) \leq \frac{4m}{\gamma^2} (O(m\kappa^2) + O(1)) = O(m^2\kappa^2) + O(m) = O(\kappa^{-2\eta}) + O(\kappa^{-(1+\eta)}),\]
where (1.7) is used. By combining (4.9) and (4.10) and letting \(\kappa\) to infinity, we derive the desired estimate. \(\square\)

Remark 4.2. Then, two conditions in (4.8) can be rewritten in terms of \(m_0\) and \(\eta\):

(i) \(\kappa > \left(\frac{8\xi m_0}{\gamma^2}\right)^{\frac{1}{2\eta}}\).
(ii) \(\gamma < \left(\frac{8\xi m_0}{\kappa^\eta}\right)^{\frac{1}{2}}\).

In (i), we only require a large coupling strength \(\kappa\) as \(\kappa \to \infty\) and there is no assumption on \(\gamma\). However for (ii), \(\gamma\) depends on \(\kappa\). In fact, \(\gamma\) also tends to zero when \(\kappa \to \infty\) as \(m\) does.

To this

5. Conclusion

In this paper, we have studied emergent dynamics of the first-order and second-order consensus models with both homogeneous and heterogeneous ensembles in which all states of the particles are restricted to the Stiefel manifold St\((p, n)\) consisting of all orthonormal \(k\)-frames in \(\mathbb{R}^n\). As mentioned in Section 1 due to the structural property of Stiefel manifolds, it has been widely used for optimization problems especially in the engineering community. Our first-order model was first suggested in [41, 42] based on the gradient flow of the total distance between all particles. For a homogeneous ensemble in which frequency matrices are the same, we show that the relaxation process towards the complete consensus state is always achieved with an exponential rate. In addition, we provide a subset of initial data leading to the complete consensus independent of the choice \((p, n)\), whereas the results in [41, 42] show that the complete consensus is globally stable if \((p, n)\) satisfies the specific relation and hence, the global consensus is achieved. On the other hand, for the heterogeneous ensemble, the phase-locked state can emerge under the large coupling strength regime, when the smallness assumption on the initial data is imposed. We also study the asymptotic
emergent dynamics of a second-order consensus model which naturally extends the first-order model by incorporating the inertial force. For a homogeneous ensemble, we present a sufficient framework leading to the complete consensus state based on a second-order Grönwall’s inequality and energy estimate. In contrast, for a heterogeneous ensemble, a sufficient framework for the practical consensus state is provided under the large coupling and small inertia regime. In fact, the phase-locked state for the second-order model is not considered in this work. Thus, we leave this problem for a future work.

**Appendix A. Proof of Lemma 3.2**

In this appendix, we provide a proof of Lemma 3.2 in which the differential inequality for \(d(S, \tilde{S})\) is derived.

Recall the relation (3.6) in Lemma 3.1

\[
\frac{d}{dt} A_{ji} = A_{ji} \Xi_i - \Xi_j A_{ji} + \frac{\kappa}{N} \sum_{k=1}^{N} a_{ik} \left[ A_{jk} - \frac{1}{2} (A_{ji} A_{ik} + A_{ji} A_{ki}) \right] + a_{jk} \left[ A_{ki} - \frac{1}{2} (A_{kj} A_{ji} + A_{jk} A_{ji}) \right]
\]

\[
= A_{ji} \Xi_i - \Xi_j A_{ji} + \frac{\kappa}{2N} \sum_{k=1}^{N} a_{ik} (A_{jk} - A_{kj}) + a_{jk} (A_{ki} - A_{ik})
\]

\[
+ \frac{\kappa}{2N} \sum_{k=1}^{N} \left( a_{ik} (A_{jk} + A_{kj}) + a_{jk} (A_{ki} + A_{ik}) - a_{ik} A_{ji} (A_{ik} + A_{ki}) - a_{jk} (A_{jk} + A_{kj}) A_{ji} \right).
\]

Similarly, we find

\[
\frac{d}{dt} \tilde{A}_{ji} = \tilde{A}_{ji} \Xi_i - \Xi_j \tilde{A}_{ji} + \frac{\kappa}{N} \sum_{k=1}^{N} a_{ik} (\tilde{A}_{jk} - \tilde{A}_{kj}) + a_{jk} (\tilde{A}_{ki} - \tilde{A}_{ik})
\]

\[
+ \frac{\kappa}{2N} \sum_{k=1}^{N} \left( a_{ik} (\tilde{A}_{jk} + \tilde{A}_{kj}) + a_{jk} (\tilde{A}_{ki} + \tilde{A}_{ik}) - a_{ik} \tilde{A}_{ji} (\tilde{A}_{ik} + \tilde{A}_{ki}) - a_{jk} (\tilde{A}_{jk} + \tilde{A}_{kj}) \tilde{A}_{ji} \right).
\]

Below, we consider the dynamics of \(A_{ji} - \tilde{A}_{ji}\):

\[
\frac{d}{dt} (A_{ji} - \tilde{A}_{ji}) = (A_{ji} - \tilde{A}_{ji}) \Xi_i - \Xi_j (A_{ji} - \tilde{A}_{ji}) + \frac{\kappa}{2N} \sum_{k=1}^{N} \mathcal{J}_{3k}
\]

\[
+ \frac{\kappa}{2N} \sum_{k=1}^{N} \left( a_{ik} ((A_{jk} - \tilde{A}_{jk}) - (A_{kj} - \tilde{A}_{kj})) + a_{jk} ((A_{ki} - \tilde{A}_{ki}) - (A_{ik} - \tilde{A}_{ik})) \right).
\]
Thus, in (A.1), we multiply 
\[ (\text{Estimate on } \Xi) \]
Then, we use the skew-symmetry of \( \Xi \)
\[ \text{identity:} \]
\[ \text{and a matrix} \]
\[ \sqrt{p} [D(S) + D(\tilde{S})] d(S, \tilde{S})^2 \]
\[ \text{and take the trace to find} \]
\[ \frac{1}{2} \frac{d}{dt} \| A_{ji} - \tilde{A}_{ji} \|^2 \]
\[ = \text{tr} \left[ \frac{\kappa}{2N} \sum_{k=1}^{N} \text{tr} [J_{3k}(A_{ji} - \tilde{A}_{ji})^\top] \right] + \frac{\kappa}{2N} \sum_{k=1}^{N} \text{tr} [J_{5k}(A_{ji} - \tilde{A}_{ji})^\top] \]
\[ + \frac{\kappa}{2N} \sum_{k=1}^{N} a_{jk} \text{tr} \left[ (A_{ki} - \tilde{A}_{ki}) (A_{ji} - \tilde{A}_{ji})^\top \right] + \frac{\kappa}{2N} \sum_{k=1}^{N} a_{jk} \text{tr} \left[ (A_{kj} - \tilde{A}_{kj}) (A_{ji} - \tilde{A}_{ji})^\top \right] \]
\[ =: J_4 + \frac{\kappa}{2N} \sum_{k=1}^{N} J_{5k} + \frac{\kappa}{2N} \sum_{k=1}^{N} J_{6k} + \frac{\kappa}{2N} \sum_{k=1}^{N} J_{7k}. \]
\[ (A.4) \]
\[ \text{Below, we estimate the terms } J_4 \text{ and } J_{5j}, \ j = 5, 6, 7, \text{ separately.} \]

- (Estimate on \( J_4 \)): for a skew-symmetric matrix \( Y \) and a matrix \( B \), we observe the following identity:
\[ \text{tr}[YBB^\top] = 0. \]

Then, we use the skew-symmetry of \( \Xi \) to see
\[ J_4 = \text{tr} \left\{ (A_{ji} - \tilde{A}_{ji}) \Xi - \Xi_j (A_{ji} - \tilde{A}_{ji})^\top \right\} \]
\[ = \text{tr} \left[ \Xi \{ (A_{ji} - \tilde{A}_{ji})^\top (A_{ji} - \tilde{A}_{ji}) \} - \text{tr} [\Xi_j (A_{ji} - \tilde{A}_{ji}) (A_{ji} - \tilde{A}_{ji})^\top] \right] \]
\[ = 0. \]

- (Estimate on \( J_{5k} \)): Due to (A.2) and (A.3), it suffices to estimate the term involving \( J_{3k,1} \). Note that
\[ \text{tr}[J_{3k,1}(A_{ji} - \tilde{A}_{ji})^\top] \leq -a_{jk} \| A_{ji} - \tilde{A}_{ji} \|^2 + a_{jk} \sqrt{p} [D(S) + D(\tilde{S})] d(S, \tilde{S})^2 \]
\[ + (a_{ik} - a_{jk}) \text{tr} ((A_{jk} - \tilde{A}_{jk}) (A_{ji} - \tilde{A}_{ji})^\top), \]
\[ \text{for } k = 1, \ldots, 4 \text{ has the same form, it suffices to provide the estimate for } J_{3k,1}: \]
\[ J_{3k,1} = a_{ik} (A_{jk} - \tilde{A}_{jk}) - a_{jk} (A_{kj} A_{ji} - \tilde{A}_{kj} \tilde{A}_{ji}) \]
\[ = a_{jk} (A_{ji} - \tilde{A}_{ji}) + a_{jk}(I_p - A_{jk}) (A_{ji} - \tilde{A}_{ji}) + a_{jk}(A_{jk} - \tilde{A}_{jk}) (I_p - A_{ji}) \]
\[ + (a_{ik} - a_{jk}) (A_{jk} - \tilde{A}_{jk}). \]
where we used the inequality:
\[ \| I_p - A_{ji} \|_F = \| S_j^T S_j - S_j^T S_i \|_F \leq \sqrt{p} D(S). \]

Hence, \( J_{5k} \) can be estimated as follows:
\[
J_{5k} \leq \text{tr}[(J_{5k,1} + \cdots + J_{5k,4})(A_{ji} - \tilde{A}_{ji})^T] \\
= -2(a_{ik} + a_{jk})\|A_{ji} - \tilde{A}_{ji}\|^2 + 2(a_{jk} + a_{ik})\sqrt{p}D(S) + D(\tilde{S})d(S, \tilde{S})^2 \\
+ (a_{ik} - a_{jk})\text{tr}((A_{jk} - \tilde{A}_{jk})(A_{ji} - \tilde{A}_{ji})^T) \\
+ (a_{ik} - a_{jk})\text{tr}((A_{kj} - \tilde{A}_{kj})(A_{ji} - \tilde{A}_{ji})^T) \\
+ (a_{jk} - a_{ik})\text{tr}((A_{ki} - \tilde{A}_{ki})(A_{ji} - \tilde{A}_{ji})^T) \\
\leq -2\kappa a_m\|A_{ji} - \tilde{A}_{ji}\|^2 + 2\kappa a_M\sqrt{p}[D(S) + D(\tilde{S})]d(S, \tilde{S})^2 + J_{5k,1},
\]
where sum of the last four terms are denoted by \( J_{5k,1} \). Then, we observe
\[
J_{5k,1} + J_{6k} + J_{7k} \\
= a_{jk}\text{tr}([(A_{ki} - \tilde{A}_{ki}) - (A_{ik} - \tilde{A}_{ik})](A_{ji} - \tilde{A}_{ji})^T) \\
+ a_{ik}\text{tr}([(A_{jk} - \tilde{A}_{jk}) - (A_{kj} - \tilde{A}_{kj})](A_{ji} - \tilde{A}_{ji})^T) \\
+ (a_{ik} - a_{jk})\text{tr}((A_{jk} - \tilde{A}_{jk})(A_{ji} - \tilde{A}_{ji})^T) \\
+ (a_{ik} - a_{jk})\text{tr}((A_{kj} - \tilde{A}_{kj})(A_{ji} - \tilde{A}_{ji})^T) \\
+ (a_{jk} - a_{ik})\text{tr}((A_{ki} - \tilde{A}_{ki})(A_{ji} - \tilde{A}_{ji})^T) \\
\leq 4(N - 1)(a_M + d(A))d(S, \tilde{S})^2.
\]
In \( \text{(A.4)} \), we collect all estimates to obtain the desired inequality:
\[
\frac{d}{dt}d(S, \tilde{S}) \leq -2\kappa(\Lambda - a_M\sqrt{p}(D(S) + D(\tilde{S})))d(S, \tilde{S}), \quad \Lambda = \left( a_m - \frac{N - 1}{N}(a_M + d(A)) \right).
\]

Appendix B. Proof of Lemma 4.3

In this appendix, we present a proof of Lemma 4.3 in which the differential inequality for \( \| D_{ij} \|_F^2 \) is derived. First, we observe the following second-order ODE of \( D_{ij} \):
\[
m\dot{D}_{ij} + \gamma D_{ij} + \kappa \xi_i D_{ij} + \kappa (\xi_i - \xi_j)S_j \\
= -m \left( S_i \dot{S}_i^T \dot{S}_i - S_j \dot{S}_j^T \dot{S}_j \right) + (S_i \Xi_i - S_j \Xi_j) \\
- \frac{m}{\gamma} \left( 2\dot{S}_i \Xi_i - S_i \Xi_i S_i^T \dot{S}_i + S_i \dot{S}_i^T S_i \Xi_i - 2\dot{S}_j \Xi_j + S_j \Xi_j S_j^T \dot{S}_j - S_j \dot{S}_j^T S_j \Xi_j \right) \\
+ \frac{\kappa}{N} \sum_{k=1}^{N} a_{jk} \left( \frac{1}{2} S_j S_j^T S_k + \frac{1}{2} S_j S_k^T S_j - S_j \right) - a_{ik} \left( \frac{1}{2} S_i S_i^T S_k + \frac{1}{2} S_i S_k^T S_i - S_i \right).
\]

(B.1)

It follows from the definition of Stiefel manifold that
\[
D_{ik}^T D_{ik} = (S_i - S_k)^T (S_i - S_k) = 2I_p - S_i^T S_k - S_k^T S_i,
\]
or equivalently,
\[
S_i - \frac{1}{2} S_i S_i^T S_k - \frac{1}{2} S_i S_k^T S_i = \frac{1}{2} S_i D_{ik}^T D_{ik}.
\]

(B.2)
In (B.1), we use the relation (B.2) to see
\[
\begin{align*}
 m\ddot{D}_{ij} + \gamma \dot{D}_{ij} + \kappa \xi_i D_{ij} + \kappa (\xi_i - \xi_j) S_j \\
 = -m(S_i \hat{S}_i^\top - S_j \hat{S}_j^\top S_j) + (S_i \hat{S}_i - S_j \hat{S}_j) + \frac{\kappa}{2N} \sum_{k=1}^N \left( a_{ik} S_i \dot{D}_{ik}^\top D_{ik} - a_{jk} S_j \dot{D}_{jk}^\top D_{jk} \right) \\
 + \frac{m}{\gamma} \left( 2\dot{S}_i \hat{S}_i - S_i \hat{S}_i S_i^\top + S_i \dot{S}_i^\top S_i \hat{S}_i - 2\dot{S}_j \hat{S}_j S_j \hat{S}_j S_j \hat{S}_j \right) \\
 =: -mI_6 + I_7 + \frac{\kappa}{2N} \sum_{k=1}^N I_{8k} + \frac{m}{\gamma} I_9.
\end{align*}
\]  
(B.3)

By using (B.3), we derive a second-order ODE of \( \|D_{ij}\|_F^2 \):
\[
\begin{align*}
 m \frac{d^2}{dt^2} \|D_{ij}\|_F^2 + \gamma \frac{d}{dt} \|D_{ij}\|_F^2 + 2\kappa \xi_i \|D_{ij}\|_F^2 + 2\kappa (\xi_i - \xi_j) \text{tr}(I_p - S_j^\top S_i) \\
 = \text{tr} \left[ m \left( \ddot{D}_{ij} D_{ij} + 2\dot{D}_{ij}^\top \dot{D}_{ij} + D_{ij}^\top D_{ij} \right) + \gamma \left( D_{ij}^\top D_{ij} + D_{ij}^\top \dot{D}_{ij} + 2\kappa D_{ij}^\top D_{ij} \right) \right] \\
 = \text{tr} \left[ (m\ddot{D}_{ij} + \gamma \dot{D}_{ij} + \kappa D_{ij})^\top D_{ij} \right] + \text{tr} \left[ D_{ij}^\top \left( m\ddot{D}_{ij} + \gamma \dot{D}_{ij} + \kappa D_{ij} \right) \right] + 2m\|\dot{D}_{ij}\|_F^2 \\
 = 2\text{tr} \left[ D_{ij}^\top \left( -mI_6 + I_7 + \frac{\kappa}{2N} \sum_{k=1}^N I_{8k} + \frac{m}{\gamma} I_9 \right) \right] + 2m\|\dot{D}_{ij}\|_F^2.
\end{align*}
\]  
(B.4)

We now assume
\[
\xi_i = \xi_j \equiv \xi, \quad i, j = 1, \cdots, N.
\]

Then, we sum the relation (B.4) with respect to \( i, j = 1, \cdots, N \) to find temporal evolutions for \( G = \frac{1}{N^2} \sum_{i,j=1}^N \|D_{ij}\|_F^2 \):
\[
\begin{align*}
 m\dddot{G} + \gamma \dot{G} + 2\kappa \xi \ddot{G} \\
 = \frac{2}{N^2} \sum_{i,j=1}^N \text{tr} \left[ D_{ij}^\top \left( -mI_6 + I_7 + \frac{\kappa}{2N} \sum_{k=1}^N I_{8k} + \frac{m}{\gamma} I_9 \right) \right] + \frac{2m}{N^2} \sum_{i,j=1}^N \|\dot{D}_{ij}\|_F^2. 
\end{align*}
\]  
(B.5)

Below, we present estimates for \( I_k, k = 6, \cdots, 9 \), respectively.

- (Estimate of \( I_6 \)): we use the fact
\[
\|PAQ^\top\|_F = \|A\|_F \quad \text{for} \quad P, Q \in \text{St}(p, n) \quad \text{and} \quad A \in \mathcal{M}_{p,p}(\mathbb{R}),
\]  
(B.6)

which can be proved as follows:
\[
\|PAQ^\top\|_F^2 = \text{tr}(QA^\top P^\top PAQ^\top) = \text{tr}(QA^\top AQ^\top) = \text{tr}(A^\top AQ^\top Q) = \text{tr}(A^\top A) = \|A\|_F^2.
\]
Then, we use (B.6) to see
\[
\text{tr} \left( D_{ij}^T I_6 \right) = \text{tr}((S_i^T - S_j^T)(S_i \dot{S}_i^T \dot{S}_i - S_j \dot{S}_j^T \dot{S}_j))
\]
\[
= \text{tr}(\dot{S}_i^T S_i - S_i^T S_j \dot{S}_j^T \dot{S}_j - S_j^T S_i \dot{S}_i^T \dot{S}_i + \dot{S}_j^T \dot{S}_j)
\]
\[
= \|\dot{S}_i\|_F^2 + \|\dot{S}_j\|_F^2 - \text{tr} \left( S_j \dot{S}_j^T \dot{S}_j^T S_i^T + S_i \dot{S}_i^T \dot{S}_i^T S_j^T \right)
\]
\[
\leq 2(\|\dot{S}_i\|_F^2 + \|\dot{S}_j\|_F^2) \leq 4D(\dot{S})^2.
\]

- (Estimate of $I_7$): we use the maximality of $\|\Xi\|_\infty$ to find
\[
\text{tr} \left( D_{ij}^T I_7 \right) = \text{tr} \left( S_i^T - S_j^T \right)(S_i \Xi - S_j \Xi) = \text{tr} \left( \Xi - S_i^T S_j \Xi - S_j^T S_i \Xi + \Xi \right)
\]
\[
\leq 2\|\Xi\|_F + 2\|\Xi\|_F \leq 4\|\Xi\|_\infty.
\]

- (Estimate of $I_8$): by straightforward calculation,
\[
\frac{1}{N^3} \sum_{i,j,k} \text{tr}(D_{ij}^T I_8k) = \frac{1}{N^3} \sum_{i,j,k} a_{ik} \text{tr}((S_i^T - S_j^T)S_i D_{ik}^T D_{ik}) - a_{jk} \text{tr}((S_i^T - S_j^T)S_j D_{jk}^T D_{jk})
\]
\[
= \frac{1}{2N^3} \sum_{i,j,k} \left( a_{ik} \text{tr}(D_{ik}^T D_{ik}) - a_{jk} \text{tr}(D_{jk}^T D_{jk})
\right.
\]
\[
+ a_{jk} S_i^T S_j D_{jk}^T D_{jk} - a_{ik} S_j^T S_i D_{ik}^T D_{ik} \right) = 0,
\]
where we used the interchange of the index $i \leftrightarrow j$.

- (Estimate of $I_9$): we use $\|S\|_F = \sqrt{p}$ for $S \in \text{St}(p,n)$ to see
\[
\text{tr} \left( D_{ij}^T I_9 \right) = \text{tr} \left[ \left( S_i^T - S_j^T \right) (2\dot{S}_i \Xi_i - S_i \Xi_i S_i^T \dot{S}_i + S_i \dot{S}_i^T S_i \Xi_i
\right.
\]
\[
- 2\dot{S}_j \Xi_j + S_j \Xi_j S_j^T \dot{S}_j - S_j \dot{S}_j^T S_j \Xi_j) \right]
\]
\[
= \text{tr}(S_i^T \dot{S}_i \Xi_i - \Xi_i S_i^T \dot{S}_i - 2S_i^T \dot{S}_j \Xi_j + S_i^T S_j \Xi_j S_j^T \dot{S}_j - S_i^T S_j \dot{S}_j^T S_j \Xi_j
\]
\[
+ S_j^T \dot{S}_j \Xi_j - \Xi_j S_j^T \dot{S}_j - 2S_j^T \dot{S}_i \Xi_i + S_j^T S_i \Xi_i S_i^T \dot{S}_i - S_j^T S_i \dot{S}_i^T S_i \Xi_i) \right)
\]
\[
\leq 8\sqrt{p}D(\dot{S})\|\Xi\|_\infty.
\]

In (B.3), we combine all estimates for $\text{tr}(D_{ij}^T I_k)$, $k = 6, \ldots, 9$ and the fact that $\|D_{ij}\|_F^2 \leq 4D(\dot{S})^2$ to obtain the desired differential inequality for $G$:
\[
mG + \gamma G + 2\kappa \xi G \leq 16mD(\dot{S})^2 + 8\|\Xi\|_\infty + \frac{16m\sqrt{p}\|\Xi\|_\infty}{\gamma} D(\dot{S}).
\]
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