A GAME OF ALIGNMENT:
COLLECTIVE BEHAVIOR OF MULTI-SPECIES

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ABSTRACT. We study the (hydro-)dynamics of multi-species driven by alignment. What distinguishes the different species is the protocol of their interaction with the rest of the crowd: the collective motion is described by different communication kernels, \( \phi_{\alpha\beta} \), between the crowds in species \( \alpha \) and \( \beta \). We show that flocking of the overall crowd emerges provided the communication array between species forms a connected graph. In particular, the crowd within each species need not interact with its own kind, i.e., \( \phi_{\alpha\alpha} = 0 \); different species which are engaged in such ‘game’ of alignment require a connecting path for propagation of information which will lead to the flocking of overall crowd. The same methodology applies to multi-species aggregation dynamics governed by first-order alignment: connectivity implies concentration around an emerging consensus.

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1. MULTI-SPECIES DYNAMICS — STATEMENT OF MAIN RESULTS

1.1. The (hydro-)dynamics of multi-species. We study the (hydro-)dynamics of multi-species driven by environmental averaging. The ‘environment’ consists of agents, each is identified by a position/velocity pair \( (x^i_{\alpha}, v^i_{\alpha}) \in (\mathbb{R}^d, \mathbb{R}^d) \). The indexing \( \{\cdot\}^i_{\alpha} \) signifies agent “\( i \)” in a species “\( \alpha \)”. What distinguishes one species from another is the way they interact with the environment: let \( \phi_{\alpha\beta} \geq 0 \) be the communication kernel between species \( \alpha \) and \( \beta \), then the dynamics describes the collective motion of agents, each of which aligns its velocity to a weighted average of velocities of neighboring agents — both from its own as well as other species,

\[
\begin{align*}
\dot{x}^i_{\alpha} &= v^i_{\alpha}, \\
\dot{v}^i_{\alpha} &= \sum_{\beta \in \mathcal{I}} \frac{1}{N^\beta} \sum_{j=1}^{N^\beta} \phi_{\alpha\beta}(|x^j_{\beta} - x^i_{\alpha}|)(v^j_{\beta} - v^i_{\alpha}),
\end{align*}
\]

subject to initial data \( (x^i_{\alpha}, v^i_{\alpha})|_{t=0} = (x^i_{\alpha_0}, v^i_{\alpha_0}) \). Here, \( N_\alpha \) is the size of the species \( \alpha \), which forms part of the overall crowd of size \( \sum_{\alpha \in \mathcal{I}} N_\alpha \). The large-crowd dynamics, \( N_{\alpha \in \mathcal{I}} \gg 1 \), is
captured by the hydrodynamic description, consult section 2,

\[
\begin{aligned}
\rho_\alpha + \nabla \cdot (u_\alpha \rho_\alpha) &= 0; \\
\rho_\alpha u_\alpha + \nabla \cdot (\rho_\alpha u_\alpha \otimes u_\alpha) &= \sum_{\beta \in \mathcal{I}} \int \phi_{\alpha\beta}(|x-y|)(u_\beta(y) - u_\alpha(x))\rho_\alpha(x)\rho_\beta(y) \, dy.
\end{aligned}
\]

(1.1)

Each of the different species is identified by a pair of density/velocity \((\rho_\alpha, u_\alpha)\), subject to initial condition \(\rho_\alpha(t, u_\alpha)|_{t=0} = (\rho_{\alpha0}, u_{\alpha0}) \in L^1_+([\mathbb{R}^d] \times W^{1,\infty}([\mathbb{R}^d]), \forall \alpha \in \mathcal{I}. \)

There are two extreme cases: when \(\phi_{\alpha\beta} \equiv \phi\) the crowd consists of a single species driven by the same communication kernel

\[
\begin{aligned}
\rho_\alpha + \nabla \cdot (\rho u) &= 0, \\
\rho u + \nabla \cdot (\rho u \otimes u) &= \int \phi(|x-y|)(u(t, y) - u(t, x))\rho(t, x)\rho(t, y) \, dy.
\end{aligned}
\]

For the large literature on the single species hydrodynamics we refer to [?] and the references therein. When \(\phi_{\alpha\beta} = \phi\delta_{\alpha\beta}\), the crowd of (1.1) splits into \(|\mathcal{I}|\) species driven by the same communication kernel, thus we end up with identical copies of the single species dynamics.

In this paper we study all the intermediate cases which involve a genuine multi-species dynamics, driven by symmetric communication array of radial decreasing kernels, \(\Phi = \{\phi_{\alpha\beta}\},\)

\[
\phi_{\alpha\beta} = \phi_{\beta\alpha} \geq 0, \quad \phi_{\alpha\beta} \text{ are radial and decreasing.}
\]

1.2. Smooth solutions must flock. Recall that the long time behavior for the single-species model is dictated by the communication kernel \(\phi\), [?, ?]: if the communication kernel \(\phi\) admits a Pareto-type ‘fat-tail” decay\(^1\), \(\phi(r) \gtrsim (1+r)^{-\theta} \) with \(\theta \leq 1\), then “smooth solutions must flock”, namely, strong solutions of the single-species model exhibit flocking behavior as

\[
\max_{x \in \text{supp } \rho_{t, x}} |u(t, x) - u_{\infty}| \xrightarrow{t \to \infty} 0.
\]

This brings us to our first main result regarding the large-time behavior of the multi-species dynamics. Let \(\Phi(r) := \{\phi_{\alpha\beta}(r)\}_{\alpha, \beta \in \mathcal{I}}\) denote the array of communication kernels associated with (1.1). The main feature here is that flocking of multi-species dynamics does not require direct, global communication among all species — we allow \(\phi_{\alpha\beta}(r)\) to vanish, indicating lack of communication between some species \(\alpha\) and \(\beta\). Instead, what matters is a minimal requirement that the communication among species forms a connected network in the sense that there is a connecting path which propagates the information of alignment between every pair of species. To this end, we introduce the weighted graph Laplacian associated with \(\Phi(r)\),

\[
\Delta_M \Phi(r)_{\alpha\beta} := \begin{cases} 
-\phi_{\alpha\beta}(r)\sqrt{M_\alpha M_\beta}, & \alpha \neq \beta; \\
\sum_{\gamma \neq \alpha} \phi_{\alpha\gamma}(r)M_\gamma, & \alpha = \beta,
\end{cases}
\]

(1.3)

where the weights, \(\{M_\alpha\}_{\alpha \in \mathcal{I}}\), are the constant masses of the different species

\[
M_\alpha(t) := \int \rho_\alpha(t, \mathbf{x}) \, d\mathbf{x} \equiv M_\alpha > 0.
\]

\(^1\)And in a slightly more general setup — if \(\phi\) is global in the sense that \(\int_0^\infty \phi(r) \, dr = \infty\).
Its properties are outlined in section 3 below. In particular, the communication array $\Phi(r)$ forms a connected graph as long as its second eigenvalue $\lambda_2(\Delta_M \Phi(r)) > 0$. Our main result shows that inter-species connectivity implies the flocking behavior of the whole crowd.

**Theorem 1.1 (Strong solutions must flock).**

Let $(\rho_\alpha(t, \cdot), u_\alpha(t, \cdot)) \in L^1_+(\mathbb{R}^d) \times W^{1,\infty}(\mathbb{R}^d)$, $\alpha \in I$ be a strong solution of the multi-species dynamics (1.1), subject to compactly supported initial conditions $(\rho_{\alpha_0}, u_{\alpha_0})$ with finite velocity fluctuations

$$\delta V_0 := \max_{\alpha,\beta \in I} \sup_{x,y \in S_0} |u_{\alpha_0}(x) - u_{\beta_0}(y)| < \infty, \quad S_0 := \cup_{\alpha} \text{supp} \{\rho_{\alpha_0}(\cdot)\}.$$ 

Then the support, $S(t) := \cup_{\alpha} \text{supp} \{\rho_\alpha(t, \cdot)\}$, remains within a finite diameter $D_\infty < \infty$ (depending on $1 - \theta$, $M$, $\delta V_0$), and the different species flock towards a limiting velocity $\mathbf{u}_\infty$,

$$\lambda_2(\Delta_M \Phi(r)) \gtrsim \frac{1}{(1 + r)^\theta}, \quad \theta < 1.$$ 

at exponential rate, $\nu$, dictated by the spatial scale $D_\infty$

$$\nu = \zeta_M \lambda_2(\Delta_M \Phi_\infty) \gtrsim \frac{\zeta_M}{(1 + D_\infty)^\theta}, \quad \Phi_\infty := \Phi(D_\infty), \quad \zeta_M := 1 - \frac{\max_\alpha M_\alpha}{\sum_\alpha M_\alpha} > 0.$$ 

The proof of theorem 1.1, carried out in section 4 below, is achieved by showing the decay of the fluctuations

$$\left( \sum_{\alpha,\beta \in I} \int |u_\alpha(t, x) - u_\beta(t, y)|^p \rho_\alpha(t, x) \rho_\beta(t, y) dx dy \right)^{1/p} \xrightarrow{t \to \infty} 0.$$ 

In particular, the decay of the energy fluctuations, corresponding to $p = 2$,

$$\delta E(t) = \sum_{\alpha,\beta \in I} \int \int |u_\alpha(t, x) - u_\beta(t, y)|^2 \rho_\alpha(t, x) \rho_\beta(t, y) dx dy,$$

and the decay of uniform fluctuations, corresponding to $p = \infty$,

$$\delta V(u(t)) = \max_{\alpha,\beta \in I} \sup_{x,y \in S(t)} |u_\alpha(t, x) - u_\beta(t, y)|, \quad S(t) = \cup_{\alpha} \text{supp} \{\rho_\alpha(t, \cdot)\},$$

imply that all species ‘aggregate’ around a limiting velocity $\mathbf{u}_\infty$. Since the total mass $M(t) = \sum_\alpha \int \rho_\alpha(t, x) dx$ and the total momentum $\mathbf{m}(t) = \sum_\alpha \int \rho_\alpha u_\alpha(t, x) dx$ are conserved in time, $M(t) = M$ and $\mathbf{m}(t) = \mathbf{m}_0$, it follows that the different species flock together with the only possible limiting velocity $\mathbf{u}_\alpha(t, \cdot) \xrightarrow{t \to \infty} \mathbf{u}_\infty := \frac{\mathbf{m}_0}{M}$.

**Remark 1.1 (Why weighted Laplacian?).** In case of equi-weighted species $M_\alpha \equiv 1$, the weighted Laplacian (1.3) amounts to the usual graph Laplacian $\Delta \Phi(r)$. Its Fiedler number, $\lambda_2(\Delta \Phi(r))$, quantifies the connectivity of the graph associated with the adjacency matrix $\Phi(r)$,
Here, we advocate the use of the weighted graph Laplacian, \( \Delta_M \Phi(r) \), whose properties are outlined in section 3 below; in particular, consult (3.9),

\[
\frac{M}{\kappa^2|\mathcal{I}|} \leq \frac{\lambda_2(\Delta_M \Phi)}{\lambda_2(\Delta \Phi)} \leq \frac{M\kappa^2}{|\mathcal{I}|}, \quad \kappa = \frac{\max_{\alpha \in \mathcal{I}} M_\alpha}{\min_{\alpha \in \mathcal{I}} M_\alpha}, \quad M := \sum_{\alpha \in \mathcal{I}} M_\alpha(0),
\]

and hence \( \Phi(r) \) is connected as long as \( \lambda_2(\Delta_M \Phi(r)) \approx_M \lambda_2(\Delta \Phi(r)) > 0 \). The advantage of using the weighted \( \lambda_2(\Delta_M \Phi(r)) \), however, is that it provides the right scaling for the decay rate of multi-species dynamics (1.5), independent of the condition number, \( \kappa \), and the number of different species, \( |\mathcal{I}| \). On the other hand, if we accept \( \kappa, |\mathcal{I}| \)-dependence, then (1.6) implies that for (1.4) to hold it suffices to verify the Pareto ‘fat-tail’ connectivity condition \( \lambda_2(\Delta \Phi(r)) \gtrsim_{M, \kappa, |\mathcal{I}|} (1 + r)^{-\theta} \) with \( \theta < 1 \).

**Remark 1.2 (Game of alignment).** The graph Laplacian of the communication array \( \Phi(r) \) is independent of the self-interacting kernels \( \{\phi_{\alpha \alpha} | \alpha \in \mathcal{I}\} \). Thus, according to theorem 1.1, flocking can be viewed as the outcome of a ‘game’ in which agents from one species interact with different species but are independent of the interaction with their own kind. Alignment dynamics based on a game within a single species was recently studied in [?]; a two-species ensemble dynamics in [?]. A main feature in our multi-species alignment game (of two or more species) is that one can ignore interactions with its own kind, i.e., set \( \phi_{\alpha \alpha} = 0 \) in (1.1) and yet the information will eventually be reflected through interactions with the other connected species leading to overall flocking.

**Example 1.1.** Consider the case of two species with \( 2 \times 2 \) symmetric communication array,

\[
\Phi = \begin{bmatrix}
0 & \phi_{12}(r) \\
\phi_{21}(r) & 0
\end{bmatrix}, \quad \phi_{12}(r) = \phi_{21}(r) \gtrsim \frac{1}{(1+r)^\theta}, \quad \theta < 1.
\]

In this case, agents in each of the two groups interact with the other group but not with their own kind \( (\phi_{11} = \phi_{22} \equiv 0) \). The large-time behavior of such ‘game’ leads to flocking. Similarity, consider the case of four species with \( 4 \times 4 \) symmetric communication array

\[
\Phi = \begin{bmatrix}
0 & \phi_{12} & 0 & \phi_{14} \\
\phi_{21} & 0 & \phi_{23} & 0 \\
0 & \phi_{32} & 0 & \phi_{34} \\
\phi_{41} & 0 & \phi_{43} & 0
\end{bmatrix}, \quad \phi_{\alpha \beta}(r) \gtrsim (1+r)^{-\mu-\min\{\alpha, \beta\}}, \quad \mu < 1/3.
\]

Again, species do not interact with their own kind, but the connectivity of inter-group interactions is strong enough to induce flocking.

We conclude by noting that the relation between connectivity and flocking was motivated by our earlier study of flocking for discrete dynamics of one species, \( \{(x^i(t), v^i(t))\}_{i=1}^N \), governed by \( \dot{v}^i = \frac{1}{N} \sum_{j=1}^N \phi(|x^j - x^i|)(v^j - v^i) \) and subject to short-range interactions, [?], Theorem 2.11. It was shown that if connectivity persists in time so that

\[
\int \lambda_2(\Delta \Phi(t))dt = \infty, \quad \Phi_{ij}(t) = \{\phi(|x_i(t) - x_j(t)|)\},
\]

then flocking follows, \( v^i(t) \to v_\infty \) as \( t \to \infty \).
1.3. One- and two-dimensional smoothness — sub-critical data. The conditional statement that ‘smooth solutions must flock’ raises the question whether the multi-species dynamics (1.1) admits global smooth solutions.

The case of one species was studied in one- and two-spatial dimensions. The one-dimensional well-posedness theory [?] provided precise characterization of global smooth solutions with sub-critical initial data, \( u'_0 + \phi \ast \rho_0 \geq 0 \). Global smoothness in two dimensions was proved for sub-critical initial data outlined in [?],[?]. Here we develop the corresponding well-posedness of multi-species dynamics (1.1) in one- and two-spatial dimensions.

The one-dimensional result is stated for non-vacuous initial data in the 1D torus.

**Theorem 1.2 (Existence of smooth solutions — one-dimensional dynamics).** Consider the multi-species dynamics (1.1) subject to non-vacuous initial data \( \{(\rho_{a0},u_{a0})\} \in (L^1_1(\mathbb{T}),W^{1,\infty}(\mathbb{T})) \). If the initial condition satisfies the sub-critical threshold condition

\[
(1.7) \quad u'_{a0}(x) + \sum_{\beta \in \mathcal{I}} \phi_{\alpha\beta} \ast \rho_{\beta0}(x) \geq 0, \quad \forall x \in \mathbb{T}, \alpha \in \mathcal{I},
\]

then the multi-species dynamics (1.1) admits global non-vacuous smooth solution.

Turning to the two-dimensional case, we let \((\rho_{a},u_{a})\) be a solution of the 2D multi-species dynamics (1.1). Global smoothness for sub-critical initial data is quantified in terms of the spectral gap associated with the (symmetric part of the) \( 2 \times 2 \) velocity gradient matrix e.g., [?]

\[
S_\alpha(t,x) := \frac{1}{2} \left( \nabla u_\alpha(t,x) + (\nabla u_\alpha(t,x))^\top \right), \quad (\nabla u_\alpha)_{ij} = \partial_j u^i_\alpha(t,\cdot), \quad i,j \in \{1,2\}.
\]

**Theorem 1.3 (Existence of smooth solutions — two-dimensional dynamics).** Consider the two-dimensional multi-species dynamics (1.1) subject to compactly supported initial conditions \( \{(\rho_{a0},u_{a0})\}_{a \in \mathcal{I}} \in (L^1_1(\mathbb{R}^2),W^{1,\infty}(\mathbb{R}^2)) \). Assume a connected communication array \( \Phi(r) = \{\phi_{\alpha\beta}(r)\}_{a,\beta \in \mathcal{I}} \) satisfying the ‘fat-tail’ decay (1.4), \( \lambda_2(\Delta \Phi(r)) \gtrsim (1 + r)^{-\theta}, \; \theta < 1 \). There exists a constant \( C_1 = C_1(|\phi'_{\alpha\beta}|_{\infty},M,\gamma) \) (specified in (5.14) below), such that if the initial fluctuations are not too large, \( \delta V_0 \leq C_1 \), and the following critical threshold conditions hold

\[
(1.8a) \quad \text{div} u_{a0}(x) + \sum_{\beta \in \mathcal{I}} \phi_{\alpha\beta} \ast \rho_{\beta0}(x) > 0, \quad \forall x \in \mathbb{R}^2,
\]

\[
(1.8b) \quad \max_{x,a} |\lambda_2(S_\alpha(0,x)) - \lambda_1(S_\alpha(0,x))| < \frac{1}{2} C_1,
\]

then the multi-species dynamics (1.1) admits a global smooth solution \((\rho_{a},u_{a}) \in C(\mathbb{R}^+_t;L^\infty \cap \mathcal{D}(\mathbb{R}^2)) \times C(\mathbb{R}^+_t;W^{1,\infty}(\mathbb{R}^2)) \) with large time hydrodynamic flocking behavior \( u_\alpha(t,x) \to \bar{u}_\alpha \).

1.4. Multi-species aggregation model. We turn our attention to the multi-species aggregation dynamics. The aggregation dynamics of a single-species arises in different contexts of modeling opinion dynamics, the rendezvous problem, etc; see e.g., [?],[?],[?],[?],[?],[?] and the reference therein,
Global smooth solutions tend to a Dirac mass which concentrates at the invariant center of mass. This large time concentration reflects the emergence of consensus (in opinion dynamics) and rendezvous problem (in distributed sensor-based dynamics) etc. There is also an increasing interest in two species-aggregation models, [7] and the recent works [7],[8], and [9]. In particular, [7],[9] study measure-valued solutions of the 2-species dynamics after blow-up and [7] categorize the possible steady states of the two-species system. Here we extend the discussion to the multi-species setting.

\begin{align}
\partial_t \rho_\alpha = -\sum_{\beta \in I} \nabla \cdot \left( (x \phi_{\alpha \beta} \ast \rho_\beta) \rho_\alpha \right) = 0, \quad \forall x \in \mathbb{R}^d, \alpha \in I.
\end{align}

The different species are identified by their densities — \( \rho_\alpha \) denotes the agent density in the species \( \alpha \), a macroscopic realization of the agent-based dynamics of a species with \( N_\alpha \) agents, each has position, \( x^i_\alpha \), and interacts with the other species

\[
\dot{x}^i_\alpha = -\sum_{\beta \in I} \frac{1}{N_\beta} \sum_{j=1}^{N_\beta} \phi_{\alpha \beta}(|x^i_\alpha - x^j_\beta|)(x^j_\beta - x^i_\alpha).
\]

In this paper, we extend the results to the multi-species setting and give explicit sufficient condition to guarantee consensus under the assumption that the communication array \( \Phi = \{ \phi_{\alpha \beta} \} \) form a connected network. Our main theorem is summarized in the following.

**Theorem 1.4 (First-order aggregation).** Let \( \{ \rho_\alpha(t, \cdot) \} \in W^1_+(\mathbb{R}^d) \) be a strong solution of the multi-species aggregation system (1.9) subject to compactly supported initial data \( (\rho_{\alpha 0})_{\alpha \in I} \) with a finite diameter

\[
D_0 = \sup_{x, y \in S_0} |x-y|, \quad S_0 = \cup_{\alpha} \text{supp} \{ \rho_{\alpha 0} \}
\]

and governed by radially symmetric decreasing kernels \( \{ \phi_{\alpha \beta}(r) \} \) (1.2). Let \( \Phi_0 \) denote the communication array scaled at the initial diameter, \( \Phi_0 = \{ \phi_{\alpha \beta}(D_0) \}_{\alpha, \beta \in I} \). There holds

\[
\delta D(t) \leq \delta D_0 \cdot e^{-2\zeta_M \lambda_2(\Delta_M \Phi_0)} t, \quad \delta D(t) := \sum_{\alpha, \beta \in I} \int \int |x - y|^2 \rho_\alpha(t, x) \rho_\beta(t, y) dx dy
\]

In particular, if the communication array \( \Phi_0 \) is connected, then the different species \( \{ \rho_\alpha \}_{\alpha \in I} \) aggregate towards the limiting position \( \bar{x}_\infty \)

\[
(1.10) \sum_{\alpha} \int |x - \bar{x}_\infty|^2 \rho_\alpha(t, x) dx \leq \sum_{\alpha} \int |x - \bar{x}_\infty|^2 \rho_{\alpha 0}(x) dx \cdot e^{-2\nu t},
\]

at exponential rate, \( \nu \), dictated by the initial spatial scale \( D_0 \),

\[
\nu = \zeta_M \lambda_2(\Delta_M \Phi_0), \quad \Phi_0 = \{ \phi_{\alpha \beta}(D_0) \}, \quad \zeta_M = 1 - \frac{\max_\alpha M_\alpha}{\sum_\alpha M_\alpha} > 0.
\]

**Remark 1.3.** The proof of theorem 1.4, carried out in section 6 below, implies that if the communication array \( \{ \phi_{\alpha \beta}(D_0) \} \) forms a connected array then all species ‘aggregate’ around a limiting position \( \bar{x}_\infty \). Since the center of mass \( \frac{1}{M(t)} \sum_\alpha \int \rho_\alpha(t, x) x dx \) is conserved in time, it follows that the different species aggregate around \( \bar{x}_\infty = \text{center of mass} \) as the only possible
limiting position. As before, aggregation depends on path connectivity but are independent of the self-interacting kernels, \( \{ \phi_{\alpha \alpha} | \alpha \in I \} \) which are allowed to vanish.

**Remark 1.4** (Existence of smooth solution). Assume that \( x \phi_{\alpha \beta} \in W^{1,\infty}(\mathbb{R}^d) \). Then, the multi-species dynamics which we rewrite as

\[
\partial_t \rho_\alpha + \sum_{\beta \in I} ((x \phi_{\alpha \beta}) \ast \rho_\beta) \cdot \nabla \rho_\alpha = - \sum_{\beta \in I} \nabla \cdot ((x \phi_{\alpha \beta}) \ast \rho_\beta) \rho_\alpha
\]

implies the uniform bound

\[
\frac{d}{dt} |\rho_\alpha|_\infty \leq \sum_{\beta \in I} |\nabla \cdot (x \phi_{\alpha \beta}) \ast \rho_\beta|_\infty |\rho_\alpha|_\infty \lesssim \sum_{\beta \in I} |\nabla \cdot (x \phi_{\alpha \beta})|_\infty M_\beta |\rho_\alpha|_\infty.
\]

The uniform bound of the \( \rho_\alpha \)'s implies higher \( H^s \) Sobolev bounds by standard energy estimates. Thus, for example we have the \( H^1 \)-bound

\[
\frac{d}{dt} \sum_{\alpha \in I} |\nabla \rho_\alpha|_2^2 = \sum_{\alpha, \beta \in I} \int |\nabla \rho_\alpha \nabla (\phi_x \ast \rho_\beta) + \phi_x \ast \rho_\beta \cdot \nabla \rho_\alpha| \, dx \\
\leq \sum_{\alpha, \beta \in I} \left( |\nabla \rho_\alpha|_2 |\nabla (\phi_x)|_\infty |\nabla \rho_\beta|_2 |\rho_\alpha|_\infty + 3 |\nabla \rho_\alpha|_2^2 |\nabla (\phi_x)|_\infty |\rho_\beta|_1 \right) \\
\lesssim \sum_{\alpha \in I} |\nabla \rho_\alpha|_2^2.
\]

The paper is organized as follows: In section 2, we formally derive the macroscopic model (1.1) as the large-crowd dynamic description of the discrete agent-based model. In section 3 we prepare the weighted Poincaré inequality associated with weighted graph Laplacian which will be used in the sequel. In section 4, we prove the main results of flocking: decay of energy fluctuations in theorem 4.1 and decay of uniform fluctuations in 4.2, which in turn lead to the proof of theorem 1.1. In section 5, we prove the existence of global smooth solutions — the one- and two-dimensional setup in theorem 1.2 and respectively 1.3. Finally in section 6, we treat the multi-species aggregation of system, proving Theorem 1.4.

### 2. Derivation of the mesoscopic and hydrodynamic models

In this section, we formally derive the multi-species hydrodynamics (1.1) from the underlying multi-species agent-based dynamics. To this end, we first derive a mesoscopic Vlasov type description which in turn yields the macroscopic description (1.1).

To formulate the mesoscopic equation, we first define the following empirical probability measure associated to the species \( \alpha \), which represents the probability of finding an agent from species \( \alpha \) at position \( x \) with velocity \( v \):

\[
f_\alpha(t, x, v) = \frac{1}{N_\alpha \sum_{i=1}^{N_\alpha}} \delta_{x_{\alpha_i}(t)} \otimes \delta_{v_{\alpha_i}(t)}.
\]

Here \( N_\alpha \) denotes the number of agents in the group \( \alpha \). Evolution of each probability density \( f_\alpha \) can be derived by testing \( \partial_t f_\alpha \) against an arbitrary smooth function \( \varphi \) through equation
\begin{align}
\int \int \partial_t f_\alpha(t, x, v) \varphi(x, v) dx dv &= \frac{1}{N_\alpha} \sum_{i=1}^{N_\alpha} \partial_t \varphi(x_\alpha^i(t), v_\alpha^i(t)) \\
&= \frac{1}{N_\alpha} \sum_{i=1}^{N_\alpha} \left[ \dot{x}_\alpha^i \cdot \nabla_x \varphi(x_\alpha^i(t), v_\alpha^i(t)) + \dot{v}_\alpha^i \cdot \nabla_v (\varphi(x_\alpha^i, v_\alpha^i)) \right] \\
&= \frac{1}{N_\alpha} \sum_{i=1}^{N_\alpha} \left[ v_\alpha^i \cdot \nabla_x \varphi(x_\alpha^i, v_\alpha^i) + F_\alpha^i \cdot \nabla_v \varphi(x_\alpha^i, v_\alpha^i) \right],
\end{align}

with an alignment forcing \( F_\alpha^i \) given by

\[ F_\alpha^i = \sum_{\beta \in \mathcal{I}} \frac{1}{N_\beta} \sum_{j=1}^{N_\beta} \phi_{\alpha\beta}(|x_\beta^j - x_\alpha^i|)(v_\beta^j - v_\alpha^i) = \sum_{\beta \in \mathcal{I}} L_{\alpha\beta}(f_\beta)(x_\alpha^i, v_\alpha^i), \]

where \( L_{\alpha\beta}(f_\beta)(x_\alpha^i, v_\alpha^i) := \int \int \phi_{\alpha\beta}(|y - x_\alpha^i|)(w - v_\alpha^i)f_\beta(y, w)dy dw \). Formal integration by parts in (2.2) yields

\begin{align}
\int \int \partial_t f_\alpha(t, x, v) \varphi(x, v) dx dv &= \int \int [v \cdot \nabla_x \varphi(x, v) + \sum_{\beta \in \mathcal{I}} L_{\alpha\beta}(f_\beta)(x, v) \cdot \nabla_v \varphi(x, v)] f_\alpha(x, v) dx dv \\
&= - \int \int \left[ v \cdot \nabla_x f_\alpha(x, v) + \nabla_v \cdot \left( \sum_{\beta \in \mathcal{I}} L_{\alpha\beta}(f_\beta) f_\alpha \right) \right] \varphi dx dv.
\end{align}

Since the test function \( \varphi \) is arbitrary, the above integral equation yields the \emph{mesoscopic scale equation}

\begin{equation}
\partial_t f_\alpha(x, v) + v \cdot \nabla_x f_\alpha(x, v) + \nabla_v \cdot \left( \sum_{\beta \in \mathcal{I}} L_{\alpha\beta}(f_\beta) f_\alpha \right) = 0.
\end{equation}

The bi-linear expression inside the parenthesis on the left represents the inter-species alignment interactions. This completes the derivation from the microscopic agent-based dynamics to the mesoscopic scale dynamics.

The hydrodynamic description is formally achieved by calculating the time evolution of the \emph{observable moments}, e.g., the mass density and the momentum density:

\begin{equation}
\begin{cases}
\rho_\alpha(t, x) := \int_{\mathbb{R}^d} f_\alpha(t, x, v) dv; \\
\rho_\alpha u_\alpha(t, x) := \int_{\mathbb{R}^d} v f_\alpha(t, x, v) dv.
\end{cases}
\end{equation}

By integrating the mesoscopic equation (2.3) in the velocity variable \( v \) and applying integration by parts, we derive the mass equation for \( \rho_\alpha \):

\begin{equation}
(\rho_\alpha)_t + \nabla_x \cdot (\rho_\alpha u_\alpha) = 0, \quad \forall \alpha \in \mathcal{I}.
\end{equation}
The evolution of the momentum \( \rho_\alpha u_\alpha \) is obtained by integrating (2.3) against \( v \),

\[
0 = \int_{\mathbb{R}^d} \left[ \partial_t (v f_\alpha) + v (v \cdot \nabla_x f_\alpha) + v \nabla_v \cdot \left( \sum_{\beta \in I} L_{\alpha \beta} (f_\beta) f_\alpha \right) \right] \, dv =: I + II + III.
\]

The first term is the time derivative of the momentum density, \( \rho_\alpha u_\alpha \) in (2.4),

\[
I = \partial_t (\rho_\alpha u_\alpha);
\]

the second term \( II \) can be rewritten as

\[
II = \nabla_x \cdot (\rho_\alpha u_\alpha \otimes u_\alpha) + \nabla_x \cdot \int_{\mathbb{R}^d} (u_\alpha - v) \otimes (u_\alpha - v) f_\alpha (x, v_\alpha) \, dv
=: \nabla_x \cdot (\rho_\alpha u_\alpha \otimes u_\alpha) + \nabla_x \cdot P_\alpha,
\]

where \( P_\alpha \) is interpreted as pressure tensor. For the third term \( III \) in (2.6), we use integration by parts to rewrite it as follows

\[
III = \int v \nabla_v \cdot \left( \sum_{\beta \in I} L_{\alpha \beta} (f_\beta) f_\alpha \right) \, dv = - \sum_{\beta \in I} \int \phi_{\alpha \beta} (|y - x|) (w - v) f_\beta (y, w) f_\alpha (x, v) \, dy \, dw \, dv
= - \sum_{\beta \in I} \int \int \phi_{\alpha \beta} (|y - x|) (w f_\beta (y, w)) f_\alpha (x, v) \, dw \, dy \, dv
\]

\[
= - \sum_{\beta \in I} \int \int \phi_{\alpha \beta} (|y - x|) (\rho_\beta u_\beta) (y) f_\alpha (x, v) \, dy \, dv
+ \sum_{\beta \in I} \int \int \phi_{\alpha \beta} (|y - x|) f_\beta (y, w) (\rho_\alpha u_\alpha) (x) \, dy \, dw
\]

\[
= - \sum_{\beta \in I} \int \phi_{\alpha \beta} (|x - y|) (u_\beta (y) - u_\alpha (x)) \rho_\alpha (x) \rho_\beta (y) \, dy.
\]

Now combining (2.7), (2.8) and (2.9) we obtain the hydrodynamic momentum equation

\[
\partial_t (\rho_\alpha u_\alpha) + \nabla \cdot (\rho_\alpha u_\alpha \otimes u_\alpha) + \nabla_x \cdot P_\alpha = \sum_{\beta \in I} \int \phi_{\alpha \beta} (|x - y|) (u_\beta (y) - u_\alpha (x)) \rho_\alpha (x) \rho_\beta (y) \, dy.
\]

Similar to the one-species (hydro-)dynamics, \([?], [?]\), we limit ourselves to the mono-kinetic ansatz \( f_\alpha (x, v) = \rho_\alpha (x) \delta_{u_\alpha (x)} (v) \) to impose the pressure closure \( P_\alpha \equiv 0 \), and end up with the multi-species hydrodynamics (1.1).
3. Weighted Poincaré inequalities

Given an $N \times N$ array $A = \{a_{\alpha \beta}\}$ of non-negative entries and $N$-vector of positive weights $\{w_\alpha\}$, we are concerned with a weighted Poincaré inequality of the form

\begin{equation}
\sum_{\alpha, \beta} a_{\alpha \beta} |x_\alpha - x_\beta|^2 w_\alpha w_\beta \geq \nu \sum_{\alpha, \beta} |x_\alpha - x_\beta|^2 w_\alpha w_\beta, \quad \nu > 0.
\end{equation}

The standard Poincaré (or Courant-Fisher) inequality tells us that, in case of equal weights $w_\alpha \equiv 1$, (3.1) holds with optimal $\nu$ given by the Fielder number, $\nu = \lambda_2(\Delta A)/N$, where $\Delta A$ is the graph Laplacian, [?], [?], proposition 6.1,

\begin{equation}
\sum_{\alpha, \beta} a_{\alpha \beta} |x_\alpha - x_\beta|^2 \geq \frac{\lambda_2(\Delta A)}{N} \sum_{\alpha, \beta} |x_\alpha - x_\beta|^2. \quad (\Delta A)_{\alpha \beta} := -(1 - \delta_{\alpha \beta})a_{\alpha \beta} + \delta_{\alpha \beta} \sum_{\gamma \neq \alpha} a_{\alpha \gamma}.
\end{equation}

To treat the case of general weights, we let $\Delta_w A$ denote the weighted Laplacian

\begin{equation}
(\Delta_w A)_{\alpha \beta} = \begin{cases} 
-a_{\alpha \beta} \sqrt{w_\alpha w_\beta}, & \alpha \neq \beta, \\
\sum_{\gamma \neq \alpha} a_{\alpha \gamma} w_\gamma, & \alpha = \beta.
\end{cases}
\end{equation}

Observe that $\Delta_w A$ is symmetric yet not row stochastic. Its second eigenvalue dictates the following weighted Poincaré inequality for arbitrary $N$-vectors $x = \{x_\alpha\}$.

Lemma 3.1 (Weighted Poincaré inequality – vectors). There holds

\begin{equation}
\sum_{\alpha, \beta} a_{\alpha \beta} |x_\alpha - x_\beta|^2 w_\alpha w_\beta \geq \frac{\lambda_2(\Delta_w A)}{\sum_\beta w_\beta} \sum_{\alpha, \beta} |x_\alpha - x_\beta|^2 w_\alpha w_\beta.
\end{equation}

Remark 3.1 (Scaling). Lemma 3.1 with $w \equiv 1$ recovers the regular Poincaré (3.2). Observe that (3.2) together with the obvious $\min w_\alpha^2 \leq w_\alpha w_\beta \leq \max w_\alpha^2$ yield a desired bound (3.1) with $\nu = \lambda_2(\Delta A)\kappa^{-2}/N$,

\begin{equation}
\sum_{\alpha, \beta} a_{\alpha \beta} |x_\alpha - x_\beta|^2 w_\alpha w_\beta \geq \lambda_2(\Delta A) \frac{1}{\kappa^2 N} \sum_{\alpha, \beta} |x_\alpha - x_\beta|^2 w_\alpha w_\beta, \quad \kappa := \max w_\alpha / \min w_\alpha.
\end{equation}

The point to note here is that this bound in terms of $\lambda_2(\Delta A)$ depends on $N$ and the condition number $\kappa$. In contrast, the weighted bound (3.4) which involves $\lambda_2(\Delta_w(A))$ has the right ‘scaling’, depending on the (usually invariant) total mass of the weights but otherwise it is independent $N, \kappa$.

Proof of Lemma 3.1. The sum on the left of (3.4) can be expressed as a bi-linear form in terms of the weighted Laplacian $\Delta_w A$ in (3.3) (here and below we abbreviate $\sqrt{w} x \equiv (\sqrt{w_1} x_1, \ldots, \sqrt{w_N} x_N)^T$)

\begin{equation}
\langle (\Delta_w A) \sqrt{w} x, \sqrt{w} x \rangle := -\sum\sum a_{\alpha \beta} \sqrt{w_\alpha w_\beta} \sqrt{w_\alpha} \sqrt{w_\beta} x_\alpha x_\beta + \sum a_{\alpha \beta} \sum_{\beta \neq \alpha} a_{\alpha \beta} w_\beta w_\alpha |x_\alpha|^2
\end{equation}

\begin{equation}
\equiv \frac{1}{2} \sum\sum a_{\alpha \beta} |x_\beta - x_\alpha|^2 w_\alpha w_\beta,
\end{equation}
which shows that the symmetric Laplacian \( \Delta_w A \) is positive semi-definite with eigenvalues \( 0 = \lambda_1 \leq \lambda_2 \leq \ldots \). Here, \( \lambda_1 \) is the zero eigenvalue associated with the eigenvector \( \sqrt{w} := (\sqrt{w_1}, \ldots, \sqrt{w_N})^T \),

\[
\left( \left( \Delta_w A \right) \sqrt{w} \right)_\alpha = - \sum_{\beta \neq \alpha} a_{\alpha \beta} \sqrt{w_\alpha} w_\beta \sqrt{w_\beta} + \sum_{\beta \neq \alpha} a_{\alpha \beta} w_\beta \sqrt{w_\alpha} \equiv 0,
\]

and hence \( \left( \Delta_w A \right) \left( \sqrt{w} \mathbf{x} \right) = 0 \) for any constant vector \( \mathbf{x} = \mathbf{x}(1, 1, \ldots, 1)^T \). In particular, for \( \mathbf{x} = \frac{\sum_{\beta} w_\beta x_\beta}{\sum_{\beta} w_\beta} \) the orthogonal complement of \( \sqrt{w} \mathbf{x} \) is given by \( \{ \sqrt{w} (\mathbf{x} - \mathbf{x}) \} \),

\[
\langle \sqrt{w} (\mathbf{x} - \mathbf{x}), \sqrt{w} \mathbf{x} \rangle = 0, \quad \mathbf{x} := \frac{\sum_{\beta} w_\beta x_\beta}{\sum_{\beta} w_\beta},
\]

hence

\[
(3.7) \quad \langle \left( \Delta_w A \right) \sqrt{w} \mathbf{x}, \sqrt{w} \mathbf{x} \rangle = \langle \left( \Delta_w A \right) \sqrt{w} (\mathbf{x} - \mathbf{x}), \sqrt{w} (\mathbf{x} - \mathbf{x}) \rangle \geq \lambda_2(\Delta_w A) \times \left| \sqrt{w} (\mathbf{x} - \mathbf{x}) \right|^2.
\]

A straightforward computation yields

\[
\left| \sqrt{w} (\mathbf{x} - \mathbf{x}) \right|^2 = \sum_{\alpha} w_\alpha |x_\alpha|^2 - 2 \sum_{\alpha} w_\alpha x_\alpha \mathbf{x} + \sum_{\alpha} w_\alpha |\mathbf{x}|^2 = \sum_{\alpha} w_\alpha |x_\alpha|^2 - \sum_{\beta} \left| \sum_{\alpha} w_\alpha x_\alpha \right|^2 \sum_{\beta} w_\beta
\]

\[
= \frac{1}{\sum_{\beta} w_\beta} \left( \sum_{\alpha, \beta} w_\alpha w_\beta |x_\alpha|^2 - \sum_{\beta} w_\beta^2 |x_\beta|^2 - \sum_{\alpha} \sum_{\beta \neq \alpha} w_\alpha w_\beta x_\alpha x_\beta \right)
\]

\[
= \frac{1}{2 \sum_{\beta} w_\beta} \left( \sum_{\alpha} \sum_{\beta \neq \alpha} w_\alpha w_\beta |x_\alpha|^2 + \sum_{\alpha} \sum_{\beta \neq \alpha} w_\alpha w_\beta |x_\beta|^2 - 2 \sum_{\alpha} \sum_{\beta \neq \alpha} w_\alpha w_\beta x_\alpha x_\beta \right)
\]

\[
= \frac{1}{2 \sum_{\beta} w_\beta} \sum_{\alpha} \sum_{\beta \neq \alpha} |x_\alpha - x_\beta|^2 w_\alpha w_\beta,
\]

and (3.4) follows from (3.6) and (3.7).

\[\text{Remark 3.2 (Optimality). The proof of Lemma 3.1 shows the optimality of the weighted Laplacian (— choose } \sqrt{w} \mathbf{x} \text{ as the second, Fiedler eigenvector of } \Delta_w A \text{), leading to a Courant-Fisher-type characterization}
\]

\[
(3.8) \quad \frac{\lambda_2(\Delta_w A)}{\sum_{\beta} w_\beta} = \min_{|\delta \mathbf{x}|_w = 1} \sum_{\alpha} \sum_{\beta \neq \alpha} a_{\alpha \beta} |x_\alpha - x_\beta|^2 w_\alpha w_\beta, \quad |\delta \mathbf{x}|_w^2 := \sum_{\alpha} \sum_{\beta \neq \alpha} |x_\alpha - x_\beta|^2 w_\alpha w_\beta.
\]

Hence, comparing this with (3.5) one concludes

\[
(3.9) \quad \frac{1}{\kappa^2 N} \lambda_2(\Delta) \leq \lambda_2(\Delta_w A) \sum_{\beta} w_\beta \leq \frac{\kappa^2}{N} \lambda_2(\Delta), \quad \kappa = \max \frac{w_\alpha}{\min w_\alpha}.
\]

The array \( A \) forms a connected graph if it has a positive Fiedler number, \( \lambda_2(\Delta_w A) > 0 \). In particular, \( A \) being a connected graph, the degree of its nodes are positive, \( \sum_{\beta \neq \gamma} a_{\alpha \beta} w_\beta > 0 \).

To quantify this statement which will be used below, we appeal to (3.4)

\[
\sum_{\alpha} \sum_{\beta \neq \alpha} a_{\alpha \beta} |x_\alpha - x_\beta|^2 w_\alpha w_\beta \geq \lambda_2(\Delta_w A) \sum_{\beta} w_\beta \sum_{\alpha} \sum_{\beta \neq \alpha} |x_\alpha - x_\beta|^2 w_\alpha w_\beta.
\]
Fix an index $\gamma$ and test the last inequality with the vector $\{x \mid x_\alpha = \begin{cases} 0 & \alpha \neq \gamma, \\ \nu & \alpha = \gamma. \end{cases}\}$, with normalization factor $\nu = \left(2 \sum_{\beta \neq \gamma} w_\beta w_\gamma \right)^{-1/2}$ so that $|\delta x|_w = 1$. The sum on the left is reduced to the $(\gamma, \beta)$-terms with $\beta \neq \gamma$, for which $|x_\gamma - x_\beta|^2 = \nu^2$ and $(\alpha, \gamma)$-terms with $\alpha \neq \gamma$ for which $|x_\alpha - x_\gamma|^2 = \nu^2$ and (3.4) amounts to $2\nu^2 \sum_{\beta \neq \gamma} a_\gamma \beta w_\gamma w_\beta \geq \frac{\lambda_2(\Delta_w A)}{\sum_{\beta} w_\beta}$ and we conclude

$$
(3.10) \quad \text{deg}_\gamma(A) := \sum_{\beta \neq \gamma} a_\gamma \beta w_\beta \geq \frac{\sum_{\beta \neq \gamma} w_\beta}{\sum_{\beta} w_\beta} \lambda_2(\Delta_w A) \geq \zeta_w \lambda_2(\Delta_w A), \quad \zeta_w = 1 - \frac{\max_{\beta} w_\beta}{\sum_{\beta} w_\beta} > 0.
$$

Next, we extend Lemma 3.1 from vectors to vector-functions, seeking an inequality of the form

$$
\sum_{\alpha, \beta} a_{\alpha \beta} \iint |u_\alpha(x) - u_\beta(y)|^2 \rho_\alpha(x) \rho_\beta(y) dx dy \geq \nu \sum_{\alpha, \beta} \iint |u_\alpha(x) - u_\beta(y)|^2 \rho_\alpha(x) \rho_\beta(y) dx dy.
$$

(3.11)

Clearly we can use $\nu = \min_{\alpha, \beta} a_{\alpha \beta}$. But there is a sharper threshold, $\nu = \nu_A$, which allows some (and in fact most) of the entries $\{a_{\alpha \beta}\}$ to vanish yet $\nu_A > 0$. In particular, $\nu_A$ is independent of the (amplitudes of the) self-interacting terms $\{a_{\alpha \alpha}\}$.

**Lemma 3.2 (Weighted Poincaré inequality – vector-functions).** Let $\{w_\gamma\}$ be weight functions with positive masses $M_\gamma = \int w_\gamma(x) dx$. There holds

$$
\sum_{\alpha, \beta} a_{\alpha \beta} \iint |u_\alpha(x) - u_\beta(y)|^2 w_\alpha(x) w_\beta(y) dx dy \geq \nu \sum_{\alpha, \beta} \iint |u_\alpha(x) - u_\beta(y)|^2 w_\alpha(x) w_\beta(y) dx dy
$$

with $\nu = \nu_A$ given by

$$
\nu_A = \lambda_2(\Delta M A) \frac{\zeta_M}{M}, \quad \zeta_M = 1 - \frac{\max_{\gamma} M_\gamma}{M}, \quad M = \sum_{\gamma} M_\gamma.
$$

The bound (3.11) is at the heart of matter: note that the self-interacting terms $\sum_{\alpha} \iint |u_\alpha(x) - u_\alpha(y)|^2 w_\alpha(x) w_\alpha(y) dx dy$ are missing on its left but present in the lower-bound on the right.

**Proof of Lemma 3.2.** Denote the average, $\overline{u}_\alpha := \frac{\int u_\alpha(x) dx}{\int w_\alpha(x) dx}$. Since $\int_x (u_\alpha(x) - \overline{u}_\alpha) w_\alpha(x) dx$ and $\int_y (u_\beta(y) - \overline{u}_\beta) w_\beta(y) dy$ vanish, we can decompose the integral on the left of (3.11)

$$
\iint |u_\alpha(x) - u_\beta(y)|^2 \rho_\alpha(x) \rho_\beta(y) dx dy
$$

$$
\equiv \iint \left( |u_\alpha(x) - \overline{u}_\alpha|^2 + |\overline{u}_\alpha - \overline{u}_\beta|^2 + |\overline{u}_\beta - u_\beta(y)|^2 \right) w_\alpha(x) w_\beta(y) dx dy.
$$
We bound each of the three integrated terms on the right. Using (3.10), the first admits the lower-bound

\[
\sum_{\alpha \neq \beta} a_{\alpha\beta} \int \int |u_\alpha(x) - \pi_\alpha|^2 w_\alpha(x) w_\beta(y) dx dy = \sum_{\alpha} \left( \sum_{\beta \neq \alpha} a_{\alpha\beta} M_\beta \right) \int |u_\alpha(x) - \pi_\alpha|^2 w_\alpha(x) dx \\
= \sum_{\alpha} \deg_\alpha(A) \int |u_\alpha(x) - \pi_\alpha|^2 w_\alpha(x) dx \\
\geq \lambda_2(\Delta_M A) \frac{\zeta_M}{M} \sum_{\alpha,\beta} \int \int |u_\alpha(x) - \pi_\alpha|^2 w_\alpha(x) w_\beta(y) dx dy.
\]

Similarly, the third integrand is lower-bounded by

\[
\sum_{\alpha \neq \beta} a_{\alpha\beta} \int \int |\pi_\alpha - \pi_\beta|^2 w_\alpha(x) w_\beta(y) dx dy = \sum_{\alpha \neq \beta} a_{\alpha\beta} |\pi_\alpha - \pi_\beta|^2 M_\alpha M_\beta \\
\geq \lambda_2(\Delta_M A) \frac{\zeta_M}{M} \sum_{\alpha,\beta} \int \int |\pi_\alpha - \pi_\beta|^2 w_\alpha(x) w_\beta(y) dx dy.
\]

Finally, by the scalar weighted Poincaré (3.4), we bound the second integrand

\[
\sum_{\alpha \neq \beta} a_{\alpha\beta} \int \int |\pi_\alpha - \pi_\beta|^2 w_\alpha(x) w_\beta(y) dx dy = \sum_{\alpha \neq \beta} a_{\alpha\beta} |\pi_\alpha - \pi_\beta|^2 M_\alpha M_\beta \\
\geq \lambda_2(\Delta_M A) \frac{\zeta_M}{M} \sum_{\alpha,\beta} \int \int |\pi_\alpha - \pi_\beta|^2 w_\alpha(x) w_\beta(y) dx dy.
\]

Adding the last three lower-bounds we end up with

\[
\sum_{\alpha \neq \beta} a_{\alpha\beta} \int \int |u_\alpha(x) - u_\beta(y)|^2 w_\alpha(x) w_\beta(y) dx dy \\
\geq \lambda_2(\Delta_M A) \frac{\zeta_M}{M} \sum_{\alpha,\beta} \int \int \left( |u_\alpha(x) - \pi_\alpha|^2 + |\pi_\alpha - \pi_\beta|^2 + |\pi_\beta - u_\beta(y)|^2 \right) \rho_\alpha(x) w_\beta(y) dx dy \\
= \lambda_2(\Delta_M A) \frac{\zeta_M}{M} \int \int |u_\alpha(x) - u_\beta(y)|^2 w_\alpha(x) w_\beta(y) dx dy,
\]

thus proving (3.11). \qed

Remark 3.3 (Alignment and de-alignment). The weighted Poincaré inequality (3.11) involves the threshold \( \nu_A = \lambda_2(\Delta_M A) \frac{\zeta_M}{M} \) which is independent of \( \{a_{\alpha\beta}\} \): if \( A \) is connected then the non-diagonal fluctuation terms dominate the self-interacting fluctuations. In fact, this means that we can add self-fluctuations with negative amplitudes: assume that

\[
\begin{cases}
    a_{\alpha\beta} \geq 0, & \alpha \neq \beta, \\
    a_{\alpha\beta} \geq -\nu_A, & \alpha = \beta,
\end{cases}
\]

then (3.11) still survives

\[
\sum_{\alpha,\beta} a_{\alpha\beta} \int \int |u_\alpha(x) - u_\beta(y)|^2 w_\alpha(x) w_\beta(y) dx dy \\
\geq \frac{1}{2} \nu_A \sum_{\alpha,\beta} \int \int |u_\alpha(x) - u_\beta(y)|^2 w_\alpha(x) w_\beta(y) dx dy, \quad \nu_A = \lambda_2(\Delta_M A) \frac{\zeta_M}{M}.
\]
4. Smooth flocking solutions must flock

In this section, we prove the main flocking statement in theorem 1.1. The key observation is that the decay of both – the energy and uniform fluctuations are dictated by the connectivity of the multi-species configuration. To this end, let $D(t)$ denote the spatial diameter of the multi-species crowd at time $t$

\begin{equation}
D(t) := \max_{x,y \in S(t)} |x - y|, \quad S(t) = \cup_{\alpha} \text{supp} \{\rho_{\alpha}(t, \cdot)\}.
\end{equation}

Then $\Phi(D(t)) = \{\phi_{\alpha\beta}(D(t))\}$ quantifies the minimal amplitude of communication between species $\alpha$ and $\beta$ at time $t$. Our first result quantifies a minimal amount of connectivity which implies the decay of energy fluctuations

\begin{equation}
\delta E(t) := \sum_{\alpha, \beta \in I} \iint |u_{\alpha}(t, x) - u_{\beta}(t, y)|^2 \rho_{\alpha}(t, x)\rho_{\beta}(t, y) dx dy.
\end{equation}

**Theorem 4.1 (Decay of energy fluctuations).**

Let $(\rho_{\alpha}(t, \cdot), u_{\alpha}(t, \cdot)) \in L^1_1(\mathbb{R}^d) \times W^{1, \infty}(\mathbb{R}^d), \alpha \in I$, be a strong solution of the multi-species dynamics (1.1), subject to initial conditions $(\rho_{\alpha 0}, u_{\alpha 0})$ with initial energy fluctuations $\delta E_0 = \delta E(0)$. Then we have the apriori bound

\begin{equation}
\delta E(t) \leq \delta E_0 \cdot \exp \left\{ -2\zeta_M \int_0^t \lambda_2(\Delta M\Phi(D(\tau))) d\tau \right\}, \quad \zeta_M = 1 - \frac{\max_\alpha M_\alpha}{\sum_\alpha M_\alpha}.
\end{equation}

In particular, if the crowd dynamics satisfies a ‘fat-tail’ connectivity condition of Pareto type (but observe the dependence on $D(r)$ in contrast to (1.4))

\begin{equation}
\lambda_2(\Delta M\Phi(D(r))) \gtrsim \frac{1}{(1 + r)^\theta}, \quad \theta < 1,
\end{equation}

then $\delta E(t)$ decays at fractional-exponential rate

\begin{equation}
\delta E(t) \lesssim \delta E_0 \cdot e^{-2\nu_1 \cdot t^{1-\theta}}, \quad \nu_1 = \frac{\zeta_M}{1 - \theta}.
\end{equation}

**Remark 4.1.** Again, we observe that while the diagonal terms in $\delta E$ on the left of (4.3) account for fluctuations within the same species, $\iint \sum_{\alpha=\beta} |u_{\alpha}(x, t) - u_{\beta}(y, t)|^2 \rho_{\alpha}\rho_{\beta} dx dy$, the upper-bound on the right of (4.3) involves $\lambda_2(\Delta M\Phi)$ which is independent of (the amplitude of) the self-interaction terms, $\{\phi_{\alpha\alpha}\}$. One learns about the behavior of its own species by its reflection through interactions with the other connected species. In fact, arguing in view of remark 3.3 we can even allow for self-interactions with de-alignment, $\phi_{\alpha\alpha} \geq -\lambda_2(\Delta M\Phi)\frac{\zeta_M}{2M}$, and yet the overall inter-species alignment will override, yielding that the crowd will align towards $\bar{u}_\infty$.

**Proof.** Since the total mass, $M = \sum_\alpha \int \rho_{\alpha}(t, x) dx$, and total momentum, $\sum_\alpha \int \rho_{\alpha}(t, x) u_{\alpha}(t, x) dx$, are conserved in time, it follows that the decay rate of the fluctuations is the same as the decay rate of the total kinetic energy,

\begin{equation}
\frac{d}{dt} \delta E(t) = 2M \frac{d}{dt} E(t), \quad E(t) := \sum_{\alpha \in I} \int \rho_{\alpha}(t, x)|u_{\alpha}(t, x)|^2 dx.
\end{equation}
A straightforward computation using the multi-species dynamics (1.1) yields
\[
\frac{d}{dt} \left( \sum_{\alpha \in I} \rho_\alpha |u_\alpha|^2 dx \right) = 2 \int \sum_{\alpha, \beta \in I} \langle \rho_\alpha u_\alpha, \phi_{\alpha \beta} (\rho_\beta u_\beta) - (\phi_{\alpha \beta} * \rho_\beta) u_\alpha \rangle dx
\]
\[
= 2 \int \sum_{\alpha, \beta \in I} \left( \langle \rho_\alpha (x) u_\alpha (x), \phi_{\alpha \beta} (|x - y|) \rho_\beta (y) u_\beta \rangle \right) dx dy
\]
\[
= 2 \int \sum_{\alpha, \beta \in I} \langle \rho_\alpha (x) u_\alpha (x), \phi_{\alpha \beta} (|x - y|) \rho_\beta (y) u_\beta \rangle \rangle dx dy
\]
\[
- \int \sum_{\alpha, \beta \in I} \left( \rho_\alpha (x) |u_\alpha (x)|^2 \phi_{\alpha \beta} (|x - y|) \rho_\beta (y) + \rho_\beta (y) |u_\beta (y)|^2 \phi_{\beta \alpha} (|x - y|) \rho_\alpha (x) \right) dx dy
\]
\[
= - \int \sum_{\alpha, \beta \in I} \phi_{\alpha \beta} (|x - y|) |u_\alpha (t, x) - u_\beta (t, y)|^2 \rho_\alpha (t, x) \rho_\beta (t, y) dx dy.
\]
Since $\phi_{\alpha \beta}$ are decreasing, $\phi_{\alpha \beta} (|x - y|) \geq \phi_{\alpha \beta} (D(t))$, hence
\[
(4.7) \quad \frac{d}{dt} E(t) \leq - \sum_{\alpha, \beta \in I} \phi_{\alpha \beta} (D(t)) \int \int |u_\alpha (t, x) - u_\beta (t, y)|^2 \rho_\alpha (t, x) \rho_\beta (t, y) dx dy.
\]
We now appeal to the vector-function version of Poincaré inequality in Lemma 3.2, obtaining\(^2\)
\[
\frac{1}{2M} \frac{d}{dt} \delta E(t) \leq - \frac{1}{2M} (\Delta M \Phi (D(t))) \frac{C_M}{\lambda_2 (\Delta M \Phi (D(t)))} \delta E(t),
\]
and the desired bound (4.3) follows. \(\square\)

The decay of energy fluctuations, $\delta E(t)$, implies decay of pointwise fluctuations
\[
\delta V (u(t)) = \max_{a, \beta \in I, x, y \in S(t)} |u_\alpha (t, x) - u_\beta (t, y)|.
\]

**Theorem 4.2 (Decay of uniform fluctuations).**

Let $(\rho_\alpha (t, \cdot), u_\alpha (t, \cdot)) \in L^1_+ (\mathbb{R}^d) \times L^{1,\infty} (\mathbb{R}^d), \alpha \in I$, be a strong solution of the multi-species dynamics (1.1), subject to initial conditions $(\rho_{0\alpha}, u_{0\alpha})$, and assume the crowd dynamics satisfies the ‘fat-tail’ connectivity condition (4.4). Then $\delta V (u(t))$ decays at fractional-exponential rate: there exist constants $C_2 = C(\max_{\alpha, \beta} \phi_{\alpha \beta} (0), M) > 0$ and $\nu_2 = \nu(\theta, M) > 0$ such that
\[
(4.8) \quad \delta V (u(t)) \lesssim C_2 \cdot \delta V_0 \cdot e^{-2 \nu_2 \cdot t^{1-\theta}}, \quad \delta V_0 = \delta V (u(0)).
\]

**Proof.** We consider the strong solution $(\rho_\alpha, u_\alpha)$ in the non-vacuous region $x, y \in S$, where the alignment terms on the right of (1.1) admit the usual commutator form \(^3\)
\[
(4.9) \quad \partial_t u_\alpha + (u_\alpha \cdot \nabla) u_\alpha = \sum_{\beta \in I} \{ \phi_{\alpha \beta} (\rho_\beta u_\beta) - (\phi_{\alpha \beta} * \rho_\beta) u_\alpha \}, \quad \forall \alpha, \beta \in I.
\]

\(^2\)To be precise, here one employs the vector statement
\[
\frac{\lambda_2 (\Delta M \Phi)}{\sum \rho_\alpha M_\alpha} = \min_{|u| = 1} \sum_{\alpha \neq \beta \in I} \Phi_{\alpha \beta} |u_\alpha - u_\beta|^2 M_\alpha M_\beta, \quad |u|^2_M = \sum_{\alpha \neq \beta \in I} |u_\alpha - u_\beta|^2 M_\alpha M_\beta, \quad u \in \mathbb{R}^d,
\]
which follows by aggregating the scalar components of (3.11) (as was done in [?], Sec 3.1)).
Arguing along the lines of [?], we first fix an arbitrary unit vector \( \mathbf{w} \in \mathbb{R}^d \) and project (4.9) onto the space spanned by \( \mathbf{w} \) to get
\[
(\partial_t + \mathbf{u}_\alpha \cdot \nabla) \langle \mathbf{u}_\alpha(t, \mathbf{x}), \mathbf{w} \rangle = \sum_{\beta \in I} \int \phi_{\alpha\beta}(|\mathbf{x} - \mathbf{y}|) (\langle \mathbf{u}_\beta(t, \mathbf{y}), \mathbf{w} \rangle - \langle \mathbf{u}_\alpha(t, \mathbf{x}), \mathbf{w} \rangle) \rho_\beta(t, \mathbf{y}) d\mathbf{y}.
\]

Now we assume that \( \langle \mathbf{u}_\alpha(t, \mathbf{x}), \mathbf{w} \rangle \) reaches a maximum value at \((\mathbf{x}(t), \alpha(t)) = (\mathbf{x}_+(t), \alpha_+(t))\) and a minimum value at \((\mathbf{x}(t), \alpha(t)) = (\mathbf{x}_-(t), \alpha_-(t))\), denoting
\[
u_+(t) := \max_{\alpha \in I} \sup_{\mathbf{x} \in S(t)} \langle \mathbf{u}_\alpha(t, \mathbf{x}), \mathbf{w} \rangle = \mathbf{u}_{\alpha_+}(t)(\mathbf{x}_+(t)).
\]

We abbreviate \( c_{\alpha\beta}(t) := \phi_{\alpha\beta}(D(t)) \) and \( \overline{u}_\beta(t) := \frac{1}{M_\beta} \int \rho_\beta \mathbf{u}_\beta(t, \mathbf{y}) d\mathbf{y} \). Direct computation of the time evolution of \( u_+(t) \) yields,
\[
\frac{d}{dt} u_+(t) = \sum_{\beta \in I} \int \phi_{\alpha_+\beta}(|\mathbf{x}_+ - \mathbf{y}|) (\langle \mathbf{u}_\beta(t, \mathbf{y}), \mathbf{w} \rangle - \langle \mathbf{u}_{\alpha_+}(t, \mathbf{x}_+), \mathbf{w} \rangle) \rho_\beta(t, \mathbf{y}) d\mathbf{y}
\]
\[
\leq \sum_{\beta \in I} c_{\alpha_+\beta} \int (\langle \mathbf{u}_\beta(t, \mathbf{y}), \mathbf{w} \rangle - \langle \mathbf{u}_+(t), \mathbf{w} \rangle) \rho_\beta(t, \mathbf{y}) d\mathbf{y}
\]
\[
= \sum_{\beta \in I} c_{\alpha_+\beta} M_\beta \langle \overline{u}_\beta(t) - \mathbf{u}_+(t), \mathbf{w} \rangle
\]
\[
= \sum_{\beta \in I} c_{\alpha_+\beta} M_\beta \langle \overline{u}_\beta(t) - \overline{u}_\infty, \mathbf{w} \rangle + \sum_{\beta \in I} c_{\alpha_+\beta} M_\beta \langle \overline{u}_\infty - \mathbf{u}_+(t), \mathbf{w} \rangle =: I + II
\]

We proceed to show that the first term is bounded by the (rapidly decaying) energy fluctuations while the second term will contribute to the pointwise fluctuations. Indeed, since
\[
c_{\alpha\beta}(t) \leq \max_{\alpha, \beta} \phi_{\alpha\beta}(D_0) =: C_\phi,
\]
and \( M_\beta(\overline{u}_\beta(t) - \overline{u}_\infty) \equiv \frac{1}{M} \sum_{\alpha} \int \int (\mathbf{u}_\beta(t, \mathbf{y}) - \mathbf{u}_\alpha(t, \mathbf{x})) \rho_\alpha(t, \mathbf{x}) \rho_\beta(t, \mathbf{y}) d\mathbf{x} d\mathbf{y} \), then by Cauchy-Schwarz we find
\[
I \leq \frac{C_\phi}{M} \sum_{\alpha, \beta} \left( \left( \int \int |\mathbf{u}_\beta(t, \mathbf{y}) - \mathbf{u}_\alpha(t, \mathbf{x})|^2 \rho_\alpha(t, \mathbf{x}) \rho_\beta(t, \mathbf{y}) d\mathbf{x} d\mathbf{y} \right)^{1/2} \left( \int \int \rho_\alpha(t, \mathbf{x}) \rho_\beta(t, \mathbf{y}) d\mathbf{x} d\mathbf{y} \right)^{1/2} \right)
\]
\[
\leq \frac{C_\phi}{M} \left( \sum_{\alpha, \beta} \int \int |\mathbf{u}_\beta(t, \mathbf{y}) - \mathbf{u}_\alpha(t, \mathbf{x})|^2 \rho_\alpha(t, \mathbf{x}) \rho_\beta(t, \mathbf{y}) d\mathbf{x} d\mathbf{y} \times \sum_{\alpha, \beta} \int \int \rho_\alpha(t, \mathbf{x}) \rho_\beta(t, \mathbf{y}) d\mathbf{x} d\mathbf{y} \right)^{1/2}
\]
\[
= C_\phi (\delta E(t))^{1/2}.
\]

On the other hand, since \( \langle \overline{u}_\infty - \mathbf{u}_+, \mathbf{w} \rangle \leq 0 \), we use the reversed lower bound (3.10)
\[
II \leq \text{deg}_{\alpha_+}(\Phi(D(t))) \langle \overline{u}_\infty - \mathbf{u}_+, \mathbf{w} \rangle \leq \zeta_M \lambda_2(\Delta_M \Phi(D(t))) \langle \overline{u}_\infty - \mathbf{u}_+ \rangle, \quad \overline{u}_\infty := \langle \overline{u}_\infty, \mathbf{w} \rangle.
\]

The last two inequalities yield
\[
\frac{d}{dt} \mathbf{u}_+(t) \leq C_\phi (\delta E(t))^{1/2} + \zeta_M \lambda_2(\Delta_M \Phi(D(t))) \langle \overline{u}_\infty - \mathbf{u}_+ \rangle;
\]
similarly, we estimate the time evolution of $u_\cdot(t) := \min \inf_{a \in \mathcal{I}} \langle u_a(t, x), w \rangle$ obtaining
\[
\frac{d}{dt} u_\cdot(t) \geq -C_\phi (\delta E(t))^{1/2} + \zeta_* \lambda_2(\Delta M \Phi(D(t))) (\bar{\mu}_\infty - u_\cdot(t)).
\]
The difference of the last two bounds yields the apriori bound on $\delta V(u(t)) := u_+(t) - u_\cdot(t)$,
\[
\frac{d}{dt} \delta V(u(t)) \leq -\zeta_* \lambda_2(\Delta M \Phi(D(t))) \cdot \delta V(u(t)) + 2C_\phi (\delta E(t))^{1/2}.
\]
Observe that $\delta V(u(t)) = \max_{a, \beta \in \mathcal{I}} \sup_{x, y \in S(t)} \langle u_a(t, x) - u_\beta(t, y), w \rangle$ is the diameter of projected velocities on arbitrary unit vector $w$. The assumed (4.4) implies that $\delta E(t)$ admits the fractional exponential decay (4.5), and we end up with,
\[
\frac{d}{dt} \delta V(u(t)) \leq -\zeta_* \lambda_2(\Delta M \Phi(D(t))) \cdot \delta V(u(t)) + 2C_\phi \cdot (\delta E_0)^{1/2} e^{-\nu_1 \cdot t^{1-\theta}}.
\]
Finally, $(\delta E_0)^{1/2} \leq M \cdot \delta V_0$ and by assumption $\lambda_2(\Delta M \Phi(D(t))) \gtrsim (1 + t)^{-\theta}$, hence (4.8) follows by integration of (4.12).

**Remark 4.2.** Revisiting (4.10) we find
\[
\frac{d}{dt} u_+(t) \leq \sum_{\beta \in \mathcal{I}} \phi_{a, \beta}(D(t)) M_\beta (\bar{u}_\beta(t) - u_+(t), w) \leq \deg_{\alpha} \langle \Phi(D(t)) \rangle \max_{\beta \in \mathcal{I}} (\bar{u}_\beta(t) - u_+(t), w),
\]
and likewise
\[
\frac{d}{dt} u_\cdot(t) \geq \zeta_* \lambda_2(\Delta M \Phi(D(t))) \min_{\beta \in \mathcal{I}} (\bar{u}_\beta(t) - u_\cdot(t), w).
\]
The difference of the last two estimates yield the apriori bound
\[
\frac{d}{dt} \delta V(u(t)) \leq \zeta_* \lambda_2(\Delta M \Phi(D(t))) \cdot \left( -\delta V(u(t)) + \delta V(\bar{\pi}(t)) \right), \quad \delta V(\bar{\pi}) := \bar{\mu}_+ - \bar{\mu}_-.
\]
Since the diameter of averaged velocities $\delta V(\bar{\pi})$ is smaller than the diameter of the velocities $\delta V(u)$, (4.13) implies that the pointwise velocity diameter does not increase
\[
\delta V(\bar{u}(t)) \leq \delta V(u(t)) \Rightarrow \delta V(u(t)) \leq \delta V_0.
\]
Note that the apriori bound (4.14) does not require any connectivity assumption; theorem 4.2 quantifies how an additional ‘fat-tail’ connectivity (4.4) enforces the fractional exponential decay of $\delta V(u(t))$.

The last two theorems still require information on the dynamic growth of the supports $S(t) = \cup_\alpha \text{supp} \{\rho_\alpha(t, \cdot)\}$, in order to access the possible growth of $D(t)$ and the corresponding decay of $\phi_{a, \beta}(D(t))$ in (4.4). Our next result provides apriori bound how on far the different species can spread out, and this enables us to quantify flocking in terms of the connectivity of $\{\phi_{a, \beta}(r)\}$, independent of the diameter dynamics. To this end, observe that according to the apriori bound (4.14), the velocities of the different species remain bounded, and hence the spatial diameter of the support of the crowd can grow at most linearly in time: indeed, tracing the particle paths $(x(t), y(t)) \in S$ yields
\[
\frac{d}{dt} D(t) \lesssim \delta V(u(t)) \Rightarrow D(t) = \max_{x,y \in S(t)} |x - y| \lesssim D_0 + \delta V_0 \cdot t.
\]
We conclude the lower-bound (recall that $\phi_{\alpha\beta}$ are decreasing) $\phi_{\alpha\beta}(D(t)) \gtrsim \phi_{\alpha\beta}(D_0 + \delta V_0 \cdot t)$. We are now ready to prove theorem 1.1.

**Proof of theorem 1.1** proceeds in three steps.

**Step #1. Fractional exponential decay.** The variational characterization of the Fiedler number (3.4), implies that $\lambda_2(\cdot)$ is an increasing function of the non-negative entries in its argument,

$$\frac{\lambda_2(\Delta_M \Phi(D(t)))}{M} = \min_{|\delta u|_M = 1} \sum_{\alpha,\beta} \phi_{\alpha\beta}(D(t)) \cdot |u_\alpha - u_\beta|^2 M_\alpha M_\beta$$

$$\geq \min_{|\delta u|_M = 1} \sum_{\alpha,\beta} \phi_{\alpha\beta}(D_0 + \delta V_0 \cdot t) \cdot |u_\alpha - u_\beta|^2 M_\alpha M_\beta$$

$$= \lambda_2(\Delta_M \Phi(D_0 + \delta V_0 \cdot t)) \quad \frac{1}{M}.$$  

(4.16)

Hence, the Pareto decay $\lambda_2(\Delta_M \Phi(r)) \gtrsim (1 + r)^{-\theta}$ assumed in (1.4) implies $\lambda_2(\Delta_M \Phi(D(t))) \gtrsim (1 + D_0 + \delta V_0 \cdot t)^{-\theta}$ and the apriori estimate (4.3) implies

$$\delta E(t) \lesssim \delta E_0 \cdot e^{-2\nu_3 \cdot t^{1-\theta}}, \quad \nu_3 := \frac{\zeta_M}{(1 - \theta) \cdot \delta V_0}.$$  

**Step #2. Finite diameter.** The Pareto-type condition (1.4) implies an improved flocking rate of full exponential rate. Indeed, the apriori bound (4.11) together with (4.16) yield

$$\frac{d}{dt} \delta V(u(t)) \lesssim -(1 + D_0 + \delta V_0 \cdot t)^{-\theta} \cdot \delta V(u(t)) + 2C_\phi \cdot \delta E_0^{1/2} e^{-\nu_3 \cdot t^{1-\theta}}.$$  

As before we use $(\delta E_0)^{1/2} \lesssim M \cdot \delta V_0$; integrating the last inequality we find that $\delta V(u(t))$ satisfies a fractional exponential decay

$$\delta V(u(t)) \lesssim \delta V_0 \cdot e^{-\nu_4 \cdot t^{1-\theta}}, \quad \nu_4 = \min \{\nu_1, \nu_3\} > 0$$

which in turn implies a bounded spatial diameter uniformly in time,

$$\frac{d}{dt} D(t)\lesssim \delta V(u(t)) \lesssim \delta V_0 \cdot e^{-\nu_4 \cdot t^{1-\theta}} \sim D(t) \lesssim D_\infty \lesssim_{1-\theta, M} D_0 + \delta V_0 < \infty.$$  

(4.17)

**Step #3. Exponential decay.** We now have a uniform lower bound on the minimal communication, $\phi_{\alpha\beta}(D_\infty) \geq \phi_{\alpha\beta}(D_\infty)$. Hence, the monotone increasing dependence of $\lambda_2(\Delta_M A)$ on the entries of $A$, consult (3.8), we have

$$\lambda_2(\Delta_M \Phi(D(t))) \geq \lambda_2(\Delta_M \Phi_\infty) > 0, \quad \Phi_\infty := \{\phi_{\alpha\beta}(D_\infty)\}.$$  

(4.18)

We revisit the energy apriori fluctuations bound (4.3), obtaining the exponential decay

$$\delta E(t) \lesssim \delta E_0 \cdot e^{-2\nu t}, \quad \nu = \zeta_M \lambda_2(\Delta_M \Phi_\infty).$$

$$\sum_\alpha \int |u_\alpha(t, x) - \bar{u}_\infty|^2 \rho_\alpha(t, x) dx = \frac{1}{2M} \delta E(t),$$

exponential flocking (1.5) follows. Moreover, revisiting the uniform fluctuations (4.8) with (4.18) yields the exponential decay

$$\max_{\alpha \in I} \sup_{x \in S(t)} |u_\alpha(t, x) - \bar{u}_\infty| \lesssim \delta V_0 \cdot e^{-\nu t}.$$  

(4.19)
5. Existence of global smooth solutions

5.1. Critical threshold in one-dimensional flocking dynamics.

Proof of Theorem 1.2. Taking spatial derivative of the momentum equation (1.1) yields

\[(\partial_t + u_\alpha \partial_x) \left( \partial_x u_\alpha + \sum_{\beta \in I} \phi_{\alpha\beta} \ast \rho_\beta \right) = -\partial_x u_\alpha \left( \sum_{\beta \in I} \phi_{\alpha\beta} \ast \rho_\beta + \partial_x u_\alpha \right), \quad \forall \alpha \in I.\]

Thus, the “e”-quantities, \(e_\alpha := \partial_x u_\alpha + \sum_{\beta} \phi_{\alpha\beta} \ast \rho_\beta\) satisfy \(\partial_t e_\alpha + \partial_x (u_\alpha e_\alpha) = 0\) and pairing it with the mass equations \(\partial_t \rho_\alpha + \partial_x (u_\alpha \rho_\alpha) = 0\) yields

\[\partial_t q_\alpha + u_\alpha \partial_x q_\alpha = 0, \quad q_\alpha := \frac{e_\alpha}{\rho_\alpha}.\]

It follows that \(q_\alpha \geq 0\) and hence \(e_\alpha \geq 0\) are invariant zones: if \(e_\alpha(t = 0, x) \geq 0\) for all \(x \in \mathbb{T}\) then

\[\partial_x u_\alpha + \sum_{\beta \in I} \phi_{\alpha\beta} \ast \rho_\beta \geq 0, \quad \forall t \geq 0.\]

Moreover, arguing along the lines of \([?, \text{sec. 3}]\)

\[\rho_\alpha + u_\alpha \partial_x \rho_\alpha = -\partial_x u_\alpha \rho_\alpha = -\left( e_\alpha - \sum_{\beta \in I} \phi_{\alpha\beta} \ast \rho_\beta \right) \rho_\alpha = -q_\alpha^2 \rho_\alpha + \rho_\alpha \sum_{\beta \in I} \phi_{\alpha\beta} \ast \rho_\beta,

and the uniform bound \(|e_\alpha/\rho_\alpha(t, \cdot)|_\infty \leq |e_\alpha/\rho_\alpha(0, \cdot)|_\infty < \infty\) reveals that \(\rho_\alpha\) remains bounded away from vacuum.

Since \(\phi_{\alpha\beta}\) are uniformly bounded, we obtain the lower bound,

\[\partial_x u_\alpha(x, t) \geq -\sum_{\beta \in I} |\phi_{\alpha\beta}|_\infty M_\beta, \quad \forall (t, x) \in (\mathbb{R}_+, \mathbb{T}), \alpha \in I.\]

On the other hand we can see directly from the equation (5.1) that \(\partial_x u_\alpha\) has an upper bound for all time. Combining this with the lower bound, we have that \(|\partial_x u_\alpha|_\infty \leq C < \infty\) for all time and the existence of strong solutions follows. \(\square\)

5.2. Critical threshold in two-dimensional flocking dynamics.

Proof of Theorem 1.4. Our purpose is to show that the derivative \(\{\partial_t u_\alpha^i\}\) are uniformly bounded. We proceed in four steps along the lines of \([?]\) for the case of two-dimensional single species dynamics.

Step #1 — the dynamics of \(\text{div} \, \mathbf{u}_\alpha + \sum_{\beta \in I} \phi_{\alpha\beta} \ast \rho_\beta\). Differentiation of (4.9) implies that the velocity gradient matrix, \((\nabla \mathbf{u}_\alpha)_ij = \partial_j u_\alpha^i\), satisfies

\[
(\nabla \mathbf{u}_\alpha)_i + \mathbf{u}_\alpha \cdot \nabla (\nabla \mathbf{u}_\alpha) + (\nabla \mathbf{u}_\alpha)^2 = -\sum_{\beta \in I} \phi_{\alpha\beta} \ast \rho_\beta \nabla \mathbf{u}_\alpha + R_\alpha,
\]

where the entries of the residual matrices

\[
(R_\alpha)_{ij} := \sum_{\beta \in I} \int \partial_j \phi_{\alpha\beta}(|x - y|)(u_\beta^i(y) - u_\alpha^i(x))\rho_\beta(y) dy,
\]
do not exceed $| (R_{\alpha})_{ij} | \leq \sum_{\beta \in I} | \phi'_{\alpha \beta} | \infty M_\beta \cdot \delta V (t)$. The entries of the residual matrix $\{ (R_{\alpha})_{ij} \}$ can be estimated using the exponentially decaying velocity fluctuations (4.19)

\begin{equation}
(5.5) \quad | (R_{\alpha})_{ij} | \leq \sum_{\beta \in I} | \phi'_{\alpha \beta} | \infty M_\beta \cdot \delta V (t) \lesssim \delta V_0 \cdot e^{-2\nu t}.
\end{equation}

The first step is to bound the divergence: taking the trace of (5.4) we find that $d_\alpha := \nabla \cdot u_\alpha$ satisfies

\begin{equation}
(5.6) \quad (\partial_t + u_\alpha \cdot \nabla) d_\alpha + \text{Tr} (\nabla u_\alpha)^2 = - \left( \sum_{\beta \in I} \phi_{\alpha \beta} \ast \rho_\beta \right) d_\alpha + \text{Tr} R_\alpha.
\end{equation}

Arguing along the lines of [?], we invoke the mass equation and obtain the following relation,

\begin{equation}
\text{Tr} R_\alpha = \sum_{\beta \in I} \phi_{\alpha \beta} \ast \nabla \cdot (\rho_\beta u_\beta) - \sum_{\beta \in I} u_\alpha \cdot \nabla \phi_\beta \ast \rho_\beta = - \left( \sum_{\beta \in I} \phi_{\alpha \beta} \ast \rho_\beta \right) - u_\alpha \cdot \nabla \left( \sum_{\beta \in I} \phi_{\alpha \beta} \ast \rho_\beta \right) = - \left( \sum_{\beta \in I} \phi_{\alpha \beta} \ast \rho_\beta \right)',
\end{equation}

where $\left( \cdot \right)'$ denotes the material derivative. Similar to the paper [?], we define the following two quantities

\begin{equation}
(5.7) \quad \nabla u_\alpha = S_\alpha + \Omega_\alpha, \quad S_\alpha = \frac{1}{2} (\nabla u_\alpha + \nabla u_\alpha^\top), \quad \Omega_\alpha := \begin{pmatrix} 0 & -\omega_\alpha \\ \omega_\alpha & 0 \end{pmatrix},
\end{equation}

where $\omega_\alpha$ is the scaled vorticity $\omega_\alpha = \frac{1}{2} (\partial_1 u_\alpha^2 - \partial_2 u_\alpha^1)$. The symmetric part $S_\alpha$ has two real eigenvalue, i.e., $\lambda_1 (S_\alpha) \leq \lambda_2 (S_\alpha)$. Next, we recall the identity relating the trace $\text{Tr} (\nabla u_\alpha)^2$ to the spectral gap, $\lambda_2 (S_\alpha) - \lambda_1 (S_\alpha) \geq 0$, [?, eq.(2.11)],

\begin{equation}
(5.8) \quad \text{Tr} (\nabla u_\alpha)^2 \equiv \frac{d_\alpha^2 + \eta_\alpha^2 - 4\omega_\alpha^2}{2}, \quad \eta_\alpha := \lambda_2 (S_\alpha) - \lambda_1 (S_\alpha) \geq 0.
\end{equation}

Expressed in terms of $\eta_\alpha$, the trace dynamics (5.6) now reads

\begin{equation}
\left( d_\alpha + \sum_{\beta \in I} \phi_{\alpha \beta} \ast \rho_\beta \right)' = \frac{1}{2} (4\omega_\alpha^2 - \eta_\alpha^2) - \frac{1}{2} d_\alpha \left( d_\alpha + 2 \sum_{\beta \in I} \phi_{\alpha \beta} \ast \rho_\beta \right).
\end{equation}

This calls for the introduction of the new “natural” variable $e_{\alpha} = d_\alpha + \sum_{\beta \in I} \phi_{\alpha \beta} \ast \rho_\beta$, satisfying

\begin{equation}
(5.9) \quad e_{\alpha}' = \frac{1}{2} \left( \left( \sum_{\beta \in I} \phi_{\alpha \beta} \ast \rho_\beta \right)^2 + 4\omega_\alpha^2 - \eta_\alpha^2 - e_{\alpha}^2 \right).
\end{equation}

Our purpose is to show that $\{ x \mid e_{\alpha} (t, x) \geq 0, \forall \alpha \in I \}$ is invariant region of the dynamics (5.4).

Step #2 — bounding the spectral gap $\eta_\alpha$. Consider the dynamics of the symmetric part of

\begin{equation}
(S_\alpha)_t + u_\alpha \cdot \nabla S_\alpha + S_\alpha^2 - \frac{\omega_\alpha^2}{4} = - \sum_{\beta \in I} \phi_{\alpha \beta} \ast \rho_\beta S_\alpha + R_{\alpha, \text{sym}}, \quad R_{\alpha, \text{sym}} = \frac{1}{2} (R_\alpha + R_\alpha^\top).
\end{equation}
The spectral dynamics of its eigenvalues $\lambda_i(S_\alpha)$ is governed by

$$
(5.10) \quad \lambda_i' + \lambda_i^2 = \omega_a^2 - \left( \sum_{\beta \in \mathcal{I}} \phi_{\alpha \beta} \ast \rho_{\beta} \right) \lambda_i + \langle s_\alpha^i, R_{\alpha, \text{sym}} s_\alpha^i \rangle
$$
driven by the orthonormal eigenpair $\{s_\alpha^1, s_\alpha^2\}$ of the symmetric $S_\alpha$. Taking the difference, we find that $\eta_a = \lambda_2(S_\alpha) - \lambda_1(S_\alpha) \geq 0$ satisfies,

$$
(5.11) \quad (\eta_a)' + e_a \eta_a = q_a, \quad q_a := \langle s_\alpha^2, R_{\alpha, \text{sym}} s_\alpha^2 \rangle - \langle s_\alpha^1, R_{\alpha, \text{sym}} s_\alpha^1 \rangle.
$$

The residual term $q_a$ is upper-bounded by the size of the entries $\{R_{\alpha,j}^i\}$ in (5.5), $|q_a(t, \cdot)|_{\infty} \leq 2 \max_{ij} |R_{\alpha,j}^i(t, \cdot)|_{\infty} \lesssim \delta V_0 \cdot e^{-2\nu t}$. Hence, as long as $\eta_a(t, \cdot)$ remains positive, the spectral gap does not exceed

$$
(5.12) \quad |\eta_a(t, x)| \leq \max_x |\eta_a(0, x)| + \text{Const.} \frac{\delta V_0}{\nu} < C_1.
$$

The first inequality on the right follows from integration of (5.11); the second follows from the assumed bound on $|\eta_a(0)| \leq \frac{1}{2} C_1$ in (1.8b), and our choice of small enough $\delta V_0 \leq C_1$, so that $\text{Const.} \frac{\delta V_0}{\nu} \leq \frac{1}{2} C_1$; the constant $C_1$ is yet to be determined.

Step #3 — The invariance of $\mathbf{e}_\alpha(t, \cdot) \geq 0$. We return to (5.9): expressed in terms of the lower bound $\sum_{\beta \in \mathcal{I}} \phi_{\alpha \beta} \ast \rho_{\beta} \geq \sum_{\beta \in \mathcal{I}} \phi_{\alpha \beta}(D_\infty)M_\beta$ we find

$$
(5.13) \quad e_\alpha' \geq \frac{1}{2} (b_\alpha^2 - e_\alpha^2), \quad b_\alpha(t, x) := \sqrt{\left( \sum_{\beta \in \mathcal{I}} \phi_{\alpha \beta}(D_\infty)M_\beta \right)^2 - \eta_\alpha^2(t, x)}.
$$

Observe that $b_\alpha$ are well-defined: we set

$$
(5.14) \quad C_1 := \frac{1}{\sqrt{2}} \min_\alpha \sum_{\beta \in \mathcal{I}} \phi_{\alpha \beta}(D_\infty)M_\beta,
$$

so that the upper-bound (5.12) implies

$$
\left( \sum_{\beta \in \mathcal{I}} \phi_{\alpha \beta}(D_\infty)M_\beta \right)^2 - \eta_\alpha^2(t, x) \geq \frac{1}{2} C_0^2 - \eta_\alpha^2(t, x) \geq \eta_\alpha(t, x) \geq \frac{1}{\sqrt{2}} C_1 > 0.
$$

Since $e_\alpha' \geq \frac{1}{2} ((c_-)^2 - e_\alpha^2) = \frac{1}{2} (c_- - e_\alpha)(c_- + e_\alpha)$, it follows that $e_\alpha$ is increasing whenever $e_\alpha \in (-c_-, c_-)$ and in particular, if $e_\alpha(0) \geq 0, \forall \alpha \in \mathcal{I}$ then $e_\alpha(t, \cdot)$ remains positive at later times. Thus, if the initial data are sub-critical in the sense that (1.8a) holds

$$
e_\alpha(0, x) = \text{div} \mathbf{u}_\alpha(0, x) + \sum_{\beta \in \mathcal{I}} \phi_{\alpha \beta} \ast \rho_{\alpha}(0, x) \geq 0, \quad \forall x \in \mathbb{R}^2,
$$

then $e_\alpha(t, \cdot) \geq 0$ and $\eta_\alpha(t, \cdot)$ remains bounded.

Step #4 — an upper-bound of $\mathbf{e}_\alpha(t, \cdot)$. The lower-bound $\mathbf{e}_\alpha \geq 0$ implies that the vorticity is bounded. Indeed, the anti-symmetric part of (5.4) yields that the vorticity $\omega_\alpha$ satisfies

$$
(5.15) \quad \omega'_\alpha + e_\alpha \omega_\alpha = \frac{1}{2} \text{Tr} J R_\alpha, \quad J = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}
$$
hence applying (5.5) yields

\begin{equation}
|\omega'_\alpha| \leq -e_\alpha |\omega_\alpha| + \frac{1}{2} |q_\alpha|, \quad |q_\alpha(t, \cdot)| \lesssim \delta V_0 \cdot e^{-2\nu t}
\end{equation}

and we end up with same upper-bound on \( \omega_\alpha \) as with \( \eta_\alpha \),

\begin{equation}
|\omega_\alpha(t, \cdot)|_\infty \leq (\omega_\alpha)_+, \quad (\omega_\alpha)_+ := \max_x |\omega_\alpha(0, x)| + \frac{1}{2} C_1.
\end{equation}

Returning to (5.9) we have

\[ e'_\alpha \leq \frac{1}{2} \left( \left( \sum_{\beta \in I} \phi_{\alpha\beta} \ast \rho_\beta \right)^2 + 4 \omega_\alpha^2 - e_\alpha^2 \right) \leq \frac{1}{2} \left( \left( \sum_{\beta \in I} |\phi_{\alpha\beta}|_\infty \omega_\beta \right)^2 + 4 (\omega_\alpha^2 - e_\alpha^2) \right), \]

which implies that \( |e_\alpha(t, \cdot)|_\infty \leq (e_\alpha)_+ < \infty \). The uniform bound on \( e_\alpha \) implies that \( \text{div} u_\alpha \) is uniformly bounded, \( |\text{div} u_\alpha| \leq |e_\alpha|_\infty + \sum_{\beta \in I} |\phi_{\alpha\beta} \ast \rho_\beta|_\infty \leq (e_\alpha)_+ + \sum_{\beta \in I} |\phi_{\alpha\beta}|_\infty M_\beta \), and together with the bound on the spectral gap (5.12), it follows that the symmetric part \( \{S_\alpha\} \) is bounded. Finally, together with the vorticity bound (5.17) it follows that \( \{\partial_j u_{a_i}^j\} \) are uniformly bounded which completes the proof. \( \square \)

6. Multi-species aggregation dynamics

In this section, we prove Theorem 1.4. We begin by letting \( \mathbf{x}_\infty(t) \) denote the center of mass at time \( t \), i.e.,

\begin{equation}
\mathbf{x}_\infty(t) := \frac{1}{M} \sum_{a \in I} \mathbf{x}_a(t), \quad \mathbf{x}_a(t) = \int_{\mathbb{R}^d} \rho_a(t, \mathbf{x}) d\mathbf{x}.
\end{equation}

The total mass \( M = \sum_a \int \rho_a(t, \mathbf{x}) d\mathbf{x} \) is conserved in time. Moreover, by the assumed symmetry of the \( \Phi = \{\phi_{\alpha\beta}\} \) array, the total first moment is also conserved in time,

\[
\frac{d}{dt} \sum_{a \in I} \int \rho_a(t, \mathbf{x}) d\mathbf{x} = - \int \sum_{a, \beta \in I} \phi_{\alpha\beta}(|\mathbf{x} - \mathbf{y}|)(\mathbf{x} - \mathbf{y}) \rho_\alpha(t, \mathbf{x}) \rho_\beta(t, \mathbf{y}) d\mathbf{x} d\mathbf{y} = 0,
\]

since the last integrand in anti-symmetric in \((\mathbf{x}, \mathbf{y})\). Hence the center of mass is invariant in time \( \mathbf{x}_\infty(t) = \mathbf{x}_\infty(0) \).

By assumption, initial densities \( \rho_a(0) \)'s are compactly supported. What distinguishes the first-order multi-species aggregation dynamics (1.9) is the fact that the diameter of this support does not increase in time, in contrast to the possible expansion (4.17) of \( D(t) \) in the second-order flocking dynamics (1.1).

**Theorem 6.1 (Uniformly bounded support).**

Consider a strong solution of (1.1), \( \{\rho_a(t, \cdot) \in W^1_+(\mathbb{R}^d), a \in I\} \), subject to compactly supported initial data \( \{\rho_{a0}\} \). Then the diameter of its support,

\[ D(t) := \sup_{\mathbf{x}, \mathbf{y} \in S(t)} |\mathbf{x} - \mathbf{y}|, \quad S(t) = \cup_a \text{supp} \{\rho_a(t, \cdot)\} \]

does not increase in time \( D(t) \leq D_0 \).

**Proof.** There are various approaches to trace the diameter \( D(t) \) for one-species dynamics, e.g., [?, ?]. Here we proceed by considering the \( p \)-weighted diameter (\( p \)-Wasserstein metric),

\[ W_p(\rho(t)) := \int \sum_{a, \beta \in I} |\mathbf{x} - \mathbf{y}|^p \rho_a(t, \mathbf{x}) \rho_\beta(t, \mathbf{y}) d\mathbf{x} d\mathbf{y}. \]
We abbreviate \( \text{d} \cdot \text{d} = \rho_\gamma(t, z) \rho_\alpha(t, x) \rho_\beta(t, y) \text{d}x \text{d}y \text{d}z \). Differentiation yields
\[
\frac{\text{d}}{\text{d}t} W_p(\rho(t)) = \iint \sum_{\alpha, \beta \in I} |x - y|^p \partial_t \rho_\alpha(t, x) \rho_\beta(t, y) \text{d}x \text{d}y
\]
\[
+ \iint \sum_{\alpha, \beta \in I} |x - y|^p \rho_\alpha(t, x) \partial_t \rho_\beta(t, y) \text{d}x \text{d}y
\]
\[
= -2 \iint \sum_{\alpha, \beta, \gamma \in I} p|x - y|^{p-2} \langle (x - y), (x - z) \rangle \phi_\alpha \gamma(|x - z|) \text{d}m_{\alpha \beta \gamma}(t, x, y, z).
\]

The convexity \(|w|^p\) which tells us that \(|w - v|^p \geq |w|^p - p|w - v|^{p-2}(w, v)\): setting \(w = x - y\) and \(v = x - z\) we find that the last integral does not exceed
\[
- \iint \sum_{\alpha, \beta, \gamma \in I} p|x - y|^{p-2} \langle (x - y), (x - z) \rangle \phi_\alpha \gamma(|x - z|) \text{d}m_{\alpha \beta \gamma}(t, x, y, z)
\]
\[
\leq \iint \sum_{\alpha, \beta, \gamma \in I} (|z - y|^p - |x - y|^p) \phi_\alpha \gamma(|x - z|) \text{d}m_{\alpha \beta \gamma}(t, x, y, z)
\]
\[
= \iint \sum_{\alpha, \beta, \gamma \in I} |z - y|^p \phi_\alpha \gamma(|x - z|) \text{d}m_{\alpha \beta \gamma}(t, x, y, z)
\]
\[
- \iint \sum_{\alpha, \beta, \gamma \in I} |x - y|^p \phi_\alpha \gamma(|x - z|) \text{d}m_{\alpha \beta \gamma}(t, x, y, z) =: I + II
\]

Now exchange \(\alpha \leftrightarrow \gamma\) and \(x \leftrightarrow z\) in \(I\) to conclude that \(I + II = 0\), hence \(W_p(\rho(t)) \leq W_p(\rho(0))\). In particular, letting \(p \uparrow \infty\) yields the desired result \(D(t) \leq D_0\).

The case \(p = 2\) deserves special attention: in this case, we can quantify the strict decay rate of \(W_2(\rho(t))\) in term of the connectivity of the communication array \(\Phi(r)\).

**Theorem 6.2 (Decay of weighted diameter).**
Consider a strong solution of (1.1), \(\{\rho_\alpha(t, \cdot) \in W^1_p(\mathbb{R}^d), \alpha \in I\}\), subject to compactly supported initial data \(\rho_{0\alpha}\) and communication array \(\Phi_0 = \{\phi_{\alpha\beta}(D_0)\}_{\alpha, \beta \in I}\). Then the weighted diameter \(\delta D(t)\) satisfies
\[
\delta D(t) \leq e^{-2\kappa \lambda_2(\Delta M \Phi_0) t} \cdot \delta D_0, \quad \delta D(t) = \sum_{\alpha, \beta \in I} \iint |x - y|^2 \rho_\alpha(t, x) \rho_\beta(t, y) \text{d}x \text{d}y.
\]

**Proof.** We begin with computing the time evolution of \(\delta D(t) = W_2(\rho(t))\) in (6.2): the special case \(p = 2\) yields, upon exchange \(x \leftrightarrow z\),
\[
\frac{\text{d}}{\text{d}t} \left( \sum_{\alpha, \beta \in I} \iint |x - y|^2 \rho_\alpha(t, x) \rho_\beta(t, y) \text{d}x \text{d}y \right)
\]
\[
= -2M \sum_{\alpha, \beta \in I} \iint \phi_{\alpha\beta}(|x - y|) \langle (x - y), 2(x - y) \rangle \rho_\beta(t, y) \rho_\alpha(t, x) \text{d}x \text{d}y.
\]

Alternatively, since the center of mass \(\sum_\alpha \int \rho_\alpha(t, x) \text{d}x\) is invariant in time, the change of the weighted diameter \(\frac{\text{d}}{\text{d}t} \delta D(t)\) equals the rate of the total second moment \(\sum_\alpha \int |x|^2 \rho_\alpha(t, x) \text{d}x\);
arguing along the lines of the proof of theorem 4.1 we find
\[
\frac{1}{2M} \frac{d}{dt} \left( \sum_{\alpha, \beta \in I} \int\int |x - y|^2 \rho_\alpha(t, x) \rho_\beta(t, y) dx dy \right) = \frac{d}{dt} \left( \sum_{\alpha \in I} \int |x|^2 \rho_\alpha(t, x) dx \right)
\]
\[
= - \sum_{\alpha, \beta \in I} \int\int \phi_{\alpha \beta}(|x - y|) \phi_{\alpha \beta}((x - y), 2x) \rho_\beta(t, y) \rho_\alpha(t, x) dx dy
\]
\[
= - \sum_{\alpha, \beta \in I} \int\int \phi_{\alpha \beta}(|x - y|) |x - y|^2 \rho_\alpha(t, x) \rho_\beta(t, y) dx dy
\]
\[
\leq - \sum_{\alpha, \beta \in I} \phi_{\alpha \beta}(D_0) \int\int |x - y|^2 \rho_\alpha(t, x) \rho_\beta(t, y) dx dy.
\]

The last step follows from \(|x - y| \leq D(t) \leq D_0\) and recalling that \(\phi_{\alpha \beta}\) are decreasing. Using the vector version of Poincaré inequality (3.11) with \((u_\alpha(x), u_\beta(y)) = (x, y)\) we conclude
\[
\frac{1}{2M} \frac{d}{dt} \sum_{\alpha, \beta \in I} \int\int |x - y|^2 \rho_\alpha(t, x) \rho_\beta(t, y) dx dy
\]
\[
= - \lambda_2(\Delta_M \Phi(D_0)) \frac{\zeta_M}{M} \sum_{\alpha, \beta \in I} \int\int |x - y|^2 \rho_\alpha(t, x) \rho_\beta(t, y) dx dy.
\]

The bound (6.3) follows.

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References

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