KINETIC THEORY OF PARTICLE INTERACTIONS MEDIATED
BY DYNAMICAL NETWORKS*

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Abstract. We provide a detailed multiscale analysis of a system of particles interacting through a dynamical network of links. Starting from a microscopic model, via the mean field limit, we formally derive coupled kinetic equations for the particle and link densities, following the approach of [P. Degond, F. Delebecque, and D. Peurichard, Math. Models Methods Appl. Sci., 26 (2016), pp. 269–318]. Assuming that the process of remodeling the network is very fast, we simplify the description to a macroscopic model taking the form of a single aggregation-diffusion equation for the density of particles. We analyze qualitatively this equation, addressing the stability of a homogeneous distribution of particles for a general potential. For the Hookean potential we obtain a precise condition for the phase transition, and, using the central manifold reduction, we characterize the type of bifurcation at the instability onset.

Key words. individual-based model, meanfield limit, Fokker–Planck, macroscopic limit, aggregation-diffusion, linear stability, phase transition

AMS subject classifications. 82C40, 82C22, 82C26, 82C31, 92C17, 37N25

DOI. 10.1137/16M1085310

1. Introduction. Cellular materials [20], mucins [7], polymers [6, 3], and social networks [17, 1] are only a few of the numerous examples of systems involving highly dynamical networks. A detailed modeling of these systems would require understanding complex chemical, biological, or social phenomena that are difficult to probe. Nevertheless, one common feature of these systems is the strong coupling between the dynamical evolution of the individual agents (cells or monomers, for instance) with that of the network mediating their interactions. The mathematical modeling of this strongly coupled dynamics is a challenging task (see, for example, [26]) but it is a necessary step toward building more complete models of complex biological or social phenomena.

The purpose of this paper is to provide a detailed multiscale analysis—from a microscopic model to a macroscopic description, and its qualitative analysis—of a system of particles interacting through a dynamical network, in a particularly simple...
INTERACTIONS MEDIATED BY DYNAMICAL NETWORKS

setting: the basic entities are just point particles with local cross-links modeled by springs that are randomly created and destructed. In the mean field limit, assuming a large number of particles and links as well as propagation of chaos, we derive coupled kinetic equations for the particle and link densities. The link density distribution provides a statistical description of the network connectivity which turns out to be quite flexible and easily generalizable to other types of complex networks. See, e.g., another application of this methodology to networks of interacting fibers in [16]. A distinctive feature of our modeling is that the agents interact only through the network, which is described explicitly; this is an important difference with the opinion dynamics model in [1], where agents may “meet” (i.e., interact) even when they are not connected through the network.

We focus on the regime where the network evolution triggered by the linking and unlinking processes happens on a very short time scale. In other words we are interested in observing dynamical networks on a long time scale compared with the typical remodeling time scale. In this regime the link density distribution becomes a local function of the particle distribution density. The latter evolves on the slow time scale through an effective equation which takes the form of an aggregation-diffusion equation, known also as the McKean–Vlasov equation [23, 14]. The applications of such an equation with different types of diffusion range from models of collective behavior of animals through granular media and chemotaxis models to self-assembly of nanoparticles; see [28, 22, 24, 9] and the references therein. In contrast to many of the aggregation-diffusion equations studied in the literature [5, 18, 13, 4] the model derived here features a compactly supported potential. This model yields a very rich behavior, depending on two main parameters describing the interaction range and the stiffness of the connecting links, that we investigate using both linear and nonlinear techniques. In particular, we identify the parameter ranges for the linear stability/instability of the spatially homogeneous steady states. Moreover, the nonlinear analysis based on the central manifold reduction [21] provides us with a characterization of the type of bifurcation that appears at the instability onset. Such bifurcations were previously studied in [14] from a “thermodynamical” point of view, i.e., by looking at the minimizers of the free energy functional; we present here a dynamical point of view and make the connection with the thermodynamical approach. In the case without diffusion, this free energy functional reduces to the interaction energy, whose minimizers have been studied in [8, 28, 12]; for numerical studies in this direction we refer to [11]. In particular, global minimizers exist provided the associated potential is H-unstable, a classical notion in statistical mechanics linked to the phase transitions in the system [19, 27]. Moreover, it was shown in [8] that the minimizers are compactly supported for potentials with certain growth conditions at infinity. Generalization of these results to the case of compactly supported attraction-repulsion potential and linear diffusion, as in the system derived here, is a purpose of future work.

The outline of the paper is the following. In the preliminaries of section 2 we introduce an individual-based model for the point particles and the network, with rules for particles dynamics and network evolution. Then, in subsection 2.2, we derive kinetic equations in a formal way following the approach from [16] developed for systems of interacting fibers, when the number of particles $N$ and the number of links $K$ tend to infinity. In particular, we will assume that the ratio $K/N$ converges to some fixed positive limit $\xi$ that might be interpreted as an averaged number of links per particle. At the level of derivation of these equations, the precise character of particle interactions is not used and so the limit equations hold for a wide range of symmetric and integrable potentials. In section 3, we further simplify the description by assuming that the process of creating/destroying links is very fast. This enables...
us to derive a macroscopic model involving only the particle density, which takes the form of an aggregation-diffusion equation. In section 4, we analyze qualitatively this macroscopic equation, addressing the stability of a homogeneous distribution of particles for a general potential, and in section 5 we address the same question for the Hookean potential, for which we obtain a precise condition for the bifurcation. Finally, in section 6 we investigate via nonlinear analysis the character of the bifurcation, both for a rectangular (nondegenerate unstable eigenvalue) and a square (degenerate unstable eigenvalue) domain. In the last part of the paper, we illustrate the criterion distinguishing between supercritical and subcritical bifurcations for the Hookean potential and make connections with the very different approach by Chayes and Panferov in [14]. Our model is intended to provide a comprehensive treatment of a dynamical interaction network in a simple setting, and it does not allow for any meaningful quantitative comparison with real systems yet. Nevertheless, it does have some qualitative implications, which we will briefly discuss.


2.1. Preliminaries. The link between two particles located at the points $X_i$ and $X_j$ can be formed if their distance is less than a given radius of interaction $R$. If this condition is met the link is created in a Poisson process with probability $\nu_N^i$; it can also be destroyed with the probability $\nu_N^j$; both of them depend on $N$—the number of the particles in the whole system. This means that within a small time interval $\Delta t$, if two particles are located sufficiently close to each other (the distance between them is less than $R$), the link can be created with probability $\nu_N^i \Delta t$. If the two particles are already connected, the link between them can be destroyed with the probability $\nu_N^j \Delta t$ (independently of the distance between the particles). The Poisson hypothesis is chosen for the sake of simplicity. When cross-linked, the particles interact with each other subject to a pairwise potential

$$ V(X_i, X_j) = U(|X_i - X_j|). $$

For the moment we do not specify the character of interactions between the particles, trying to keep our derivation on a maximally general level.

We will first characterize the system of fixed number of particles, denoted by $N$, and fixed number of links, denoted by $K$. The equation of motion for each individual particle in the so-called overdamped regime, between two linking/unlinking events, is

$$ dX_i = -\mu \nabla X_i W dt + \sqrt{2D} dB_i, \quad i = 1, \ldots, N. $$

Above, $B_i$ is a two-dimensional Brownian motion $B_i = (B^1_i, B^2_i)$ with a positive diffusion coefficient $D > 0$, $\mu > 0$ is the mobility coefficient, and $W$ denotes the energy related to the maintenance of the links related to the potential $V$ as follows:

$$ W = \sum_{k=1}^K V(X_{i(k)}, X_{j(k)}), $$

where $i(k), j(k)$ denote the indexes of particles connected by the link $k$. Plugging this definition into expression (2), we obtain

$$ dX_i = -\mu \sum_{k=1: i(k)=i}^K \left[ \nabla x_i V(X_{i(k)}, X_{j(k)}) + \nabla x_j V(X_{i(k)}, X_{j(k)}) \right] dt + \sqrt{2D} dB_i $$

$$ = -\mu \sum_{k=1}^K [\delta_{i(k)}(i) \nabla x_i V(X_{i(k)}, X_{j(k)}) + \delta_{j(k)}(i) \nabla x_j V(X_{i(k)}, X_{j(k)})] dt $$

$$ + \sqrt{2D} dB_i. $$
Our ultimate aim is to describe the systems of large numbers of particles. From the point of view of numerical simulations, the system of N SDEs (2) for large N, although fundamental, is too complex and thus costly to handle; it is also difficult to get a qualitative understanding of the behavior of particles from (2). Therefore, in the next section we look for a “kinetic” description using probability distribution of particles and links rather than certain positions of each of the particles and links at a given time.

2.2. Derivation of the kinetic model. We introduce the empirical distributions of the particles $f^N(x,t)$ and of the links $g^K(x_1,x_2,t)$, when the numbers of particles and links are finite and equal N and K, respectively. They are equal to

\[ f^N(x,t) = \frac{1}{N} \sum_{i=1}^{N} \delta_{X_i}(x); \]

\[ g^K(x_1,x_2,t) = \frac{1}{2K} \sum_{k=1}^{K} \left[ \delta_{X_{i(k)}X_{j(k)}}(x_1,x_2) + \delta_{X_{j(k)}X_{i(k)}}(x_1,x_2) \right], \]

where the symbol $\delta_{X_i}(x)$ is the Dirac delta centered at $X_i(t)$, with a similar definition for the two-point distribution. The above measures contain the full information about the positions of particles and links at time t.

**Remark 2.1.** $g^K$ is directly related to the adjacency matrix of the underlying network $(A_{ij})_{i,j=1}^{N}$, through the equation

\[ g^K(x_1,x_2,t) = \frac{1}{2K} \sum_{i,j=1}^{N} A_{ij} \delta(x_1 - X_i) \delta(x_2 - X_j). \]

For the sake of completeness we also introduce the two-particle empirical distribution

\[ h^N(x_1,x_2,t) = \frac{1}{N(N-1)} \sum_{i \neq j} \delta_{X_i(t),X_j(t)}(x_1,x_2). \]

Obviously, the two distributions $h^N$ and $g^K$ are different, because not every pair of points is connected by a link.

The first part of this article is concerned with the derivation of the kinetic model obtained from (2) in the mean-field limit. This process is roughly speaking a derivation of equations for the limit distributions $f$ and $g$, obtained from $f^N$ and $g^K$, by letting $N$ and $K$ to infinity, i.e.,

\[ f(x,t) := \lim_{N \to \infty} f^N(x,t), \quad g(x_1,x_2,t) = \lim_{K \to \infty} g^K(x_1,x_2,t). \]

The purpose of this section is to derive the equations for evolutions of particle and links distributions $f$ and $g$ in the limit of large number of particles and fibers. We have the following formal theorem.

**Theorem 2.2.** The kinetic system

\[
\begin{align*}
\partial_t f(x,t) &= D \Delta_x f(x,t) + 2\mu \xi \nabla_x \cdot F(x,t), \\
\partial_t g(x_1,x_2,t) &= D \left( \Delta_{x_1} g(x_1,x_2,t) + \Delta_{x_2} g(x_1,x_2,t) \right) \\
&\quad + 2\mu \xi \left( \nabla_{x_1} \cdot \left( \frac{g(x_1,x_2)}{f(x_1)} F(x_1,t) \right) + \nabla_{x_2} \cdot \left( \frac{g(x_1,x_2)}{f(x_2)} F(x_2,t) \right) \right) \\
&\quad + \frac{\nu_f}{2\xi} h(x_1,x_2,t) \chi_{|x_1-x_2| \leq R} - \nu_d g(x_1,x_2,t),
\end{align*}
\]

\[ (6) \]
where

\[ F(x, t) = \int g(x, y, t) \nabla_x V(x, y) dy, \]

and

\[ f(x, t) := \lim_{N \to \infty} f_N(x, t), \]
\[ g(x_1, x_2, t) = \lim_{K \to \infty} g^K(x_1, x_2, t), \]
\[ h(x_1, x_2, t) = \lim_{K \to \infty} h^K(x_1, x_2, t), \]
\[ \nu_f = \lim_{N \to \infty} \nu_f^N(N - 1), \]
\[ \nu_d = \lim_{N \to \infty} \nu_d^N, \]

is a formal limit of the particle system (2) as \( N, K \to \infty \), provided that

\[ \lim_{K, N \to \infty} K N = \xi > 0. \]

Proof. The strategy of the proof is to first derive the equations for distribution of the particles \( f_N(x, t) \) and of the links \( g^K(x_1, x_2, t) \) in the situation when the number of each is finite and equal to \( N \) and \( K \). This happens between two linking/unlinking events in the time interval \( (t, t + \Delta t) \). We will consider the behavior of the system in this interval first and come back to the issue of creation of the new and destruction of the old links in the end of the proof.

Step 1. Let us first introduce the notation that will allow us to identify both \( f \) and \( g \) with certain distributions. Following [16, Appendix A] we first introduce the one-particle and two-particle compactly supported observable functions, \( \Phi(x) \) and \( \Psi(x_1, x_2) \), respectively, and the corresponding weak formulations \( \langle f_N(x, t), \Phi(x) \rangle \) and \( \langle g^K(x_1, x_2, t), \Psi(x_1, x_2) \rangle \) for equations of \( f_N \) and \( g^K \) (4).

Step 2. We derive the equation for the distribution of particles. Taking the time derivative of \( \langle f_N(x, t), \Phi(x) \rangle \) we get

\[ \frac{d}{dt} \langle f_N(x, t), \Phi(x) \rangle = \frac{1}{N} \sum_{i=1}^{N} \frac{d}{dt} \Phi(X_i(t)). \]

We expand this equation using (3) and Itô’s formula. Since \( dB_i \)'s are pairwise independent and are independent of \( \nabla_x \Phi(X_i(t)) \), thus assuming, for instance, that the test functions have bounded derivatives, we obtain for large \( N \)

\[ \frac{d}{dt} \langle f_N(x, t), \Phi(x) \rangle = -\frac{\mu}{N} \sum_{i=1}^{N} \nabla_x \Phi(X_i(t)) \]
\[ \times \sum_{k=1}^{K} \left[ \delta_{i(k)}(i) \nabla_{x_1} V(X_{i(k)}, X_{j(k)}) + \delta_{j(k)}(i) \nabla_{x_2} V(X_{i(k)}, X_{j(k)}) \right] \]
\[ + D \frac{1}{N} \sum_{i=1}^{N} \Delta \Phi(X_i). \]

Exchanging the order of the sums with respect to \( i \) and \( k \), using the symmetry of potential \( V \), and integration by parts we get
\[
\frac{d}{dt} \langle f^N(x,t), \Phi(x) \rangle = \frac{2\mu K}{N} \left< \left< \nabla_{x_1} \cdot \left( g^K(x_1,x_2) \nabla_{x_1} V(x_1,x_2) \right) , \Phi(x_1) \right> \right> \\
+ D \left< \left< \Delta f^N , \Phi \right> \right> \\
= \frac{2\mu K}{N} \int \nabla_{x_1} \cdot \left( \int g^K(x_1,x_2) \nabla_{x_2} V(x_1,x_2) dx_2 \right) \Phi(x_1) dx_1 \\
+ D \int \Delta f^N(x_1) \Phi(x_1) dx_1.
\]

(7)

Letting \( N, K \) to infinity, assuming that \( \frac{K}{N} \rightarrow \xi \) and that there exist the limits

\[
\lim_{N \rightarrow \infty} f^N = f \quad \text{and} \quad \lim_{K \rightarrow \infty} g^K = g,
\]

we obtain (after change of variables \( x_1 \rightarrow x, x_2 \rightarrow x' \)) a distributional formulation of equation for \( f \). The differential form of this equation is

\[
\partial_t f(x,t) = 2\mu \xi \nabla_x \cdot F(x,t) + D\Delta f, \quad F(x_1,t) = \int g(x_1,x_2,t) \nabla_x V(x_1,x_2) dx_2.
\]

Step 3. After deriving the equation for distribution of particles \( f \) we want to derive the equation for \( g \) in the analogous way. We remark that the noise in (3) transforms directly into a linear diffusion term for \( f \), and all other contributions vanish in the large \( N \) limit. It is not difficult to see that the same simplification takes place for \( g^K \) in the \( K \rightarrow \infty \) limit. Thus, to reduce the computations we will first use (3) without noise and reintroduce the diffusion term in the end.

Taking the time derivative of \( \left< \left< g^K(x_1,x_2,t), \Psi(x_1,x_2) \right> \right> \), we obtain

\[
\frac{d}{dt} \langle \langle g^K(x_1,x_2,t), \Psi(x_1,x_2) \rangle \rangle \\
= \frac{1}{2K} \sum_{k=1}^{K} \left[ \nabla_{x_1} \Psi(X_i(k),X_j(k)) \cdot \frac{d}{dt} X_i(k) + \nabla_{x_1} \Psi(X_j(k),X_i(k)) \cdot \frac{d}{dt} X_j(k) \right] \\
+ \frac{1}{2K} \sum_{k=1}^{K} \left[ \nabla_{x_2} \Psi(X_i(k),X_j(k)) \cdot \frac{d}{dt} X_j(k) + \nabla_{x_2} \Psi(X_j(k),X_i(k)) \cdot \frac{d}{dt} X_i(k) \right] \\
= E_1 + E_2.
\]

(9)

We now present how to treat \( E_1 \); \( E_2 \) can be handled analogously. We first use (3) (without noise) and transformations similar to the ones used in Step 2 to obtain

\[
E_1 = \frac{-\mu}{2K} \sum_{k'=1}^{K} \left\{ \nabla_{x_1} V(X_i(k'),X_j(k')) \times \sum_{k=1}^{K} [\delta_{i(k')}(i(k)) \nabla_{x_1} \Psi(X_i(k),X_j(k)) \\
+ \delta_{i(k')}(j(k)) \nabla_{x_1} \Psi(X_j(k),X_i(k))] \right\} \\
\times \frac{-\mu}{2K} \sum_{k'=1}^{K} \left\{ \nabla_{x_1} V(X_j(k'),X_i(k')) \times \sum_{k=1}^{K} [\delta_{j(k')}(i(k)) \nabla_{x_1} \Psi(X_i(k),X_j(k)) \\
+ \delta_{j(k')}(j(k)) \nabla_{x_1} \Psi(X_j(k),X_i(k))] \right\}.
\]

(10)
We see that the first sum with respect to $k$ in \(10\), i.e.,
\[
\sum_{k=1}^{K} \left[ \delta_{i(k)}(i(k)) \frac{\partial \Psi(X_{i(k)}, X_{j(k)})}{\partial x_1} + \delta_{i(k)}(j(k)) \frac{\partial \Psi(X_{j(k)}, X_{i(k)})}{\partial x_1} \right],
\]
does not vanish if either $i(k) = i(k')$ or $j(k) = i(k')$. To understand it better let us look at the link number $k'$. Its beginning is $i(k')$ and it is a certain fixed particle, as was the link.

If we now compute the above sum neglecting the Kronecker symbols we get $2K$ different elements. But for the Kronecker symbols included we act in the following way: we take the first link $k = 1$ and check if $i(1) = i(k')$; if yes, then definitely $j(1) \neq i(k')$ and thus the first element of the sum is equal to $\frac{\partial \Psi(X_{i(1)}, X_{j(1)})}{\partial x_1}$, but if $i(1) \neq i(k')$, then we check if $j(1) = i(k')$, and if yes the first element of the sum equals $\frac{\partial \Psi(X_{j(1)}, X_{i(1)})}{\partial x_1}$. Finally, if $i(k') \neq i(1)$ and $i(k') \neq j(1)$, the above sum reduces to the subset $k \geq 2$. Hence the maximal number of elements of the above sum is $K$, but in fact it will be equal to the number of links connected to $i(k')$ and it may be less than the number of all links $K$.

We now introduce a number of links connected to $i(k')$,
\[
C_{i(k')} = \# \{ k \mid i(k) = i(k') \text{ or } j(k) = i(k') \}.
\]
Thus, dividing \(11\) by $C_{i(k')}$ and letting $K \to \infty$ gives rise to a certain probability associated with $i(k')$, we have
\[
\lim_{K \to \infty} \frac{1}{C_{i(k')}} \sum_{k=1}^{K} \left[ \delta_{i(k')}(i(k)) \frac{\partial \Psi(X_{i(k)}, X_{j(k)})}{\partial x_1} + \delta_{i(k')}(j(k)) \frac{\partial \Psi(X_{j(k)}, X_{i(k)})}{\partial x_1} \right]
\]
\[
= 2 \int \left( \frac{\partial \Psi P}{\partial x_1} \right)(X_{i(k')}, x_2) dx_2,
\]
where
\[
P(X_{i(k')}, x_2) = \frac{g(X_{i(k')}, x_2)}{\int g(X_{i(k')}, x_2) dx_2}
\]
is a conditional probability of finding a link, provided one of its ends is at $X_{i(k')}$. We can also estimate the limit of the mean number of links per particle when $N, K \to \infty$, $\frac{K}{N} \to \xi$ more directly. Around the point $X_{i(k')}$ we have
\[
C_{i(k')} = \frac{K \int g(X_{i(k')}, x_2) dx_2}{N f^N(X_{i(k')})},
\]
therefore
\[
\lim_{K \to \infty, \frac{K}{N} \to \xi} C_{i(k')} = \frac{\xi \int g(X_{i(k')}, x_2) dx_2}{f(X_{i(k')})}.
\]
Combining \(12\) and \(13\), we obtain
\[
\lim_{N, K \to \infty, \frac{K}{N} \to \xi} \sum_{k=1}^{K} \left[ \delta_{i(k')}(i(k)) \frac{\partial \Psi(X_{i(k)}, X_{j(k)})}{\partial x_1} + \delta_{i(k')}(j(k)) \frac{\partial \Psi(X_{j(k)}, X_{i(k)})}{\partial x_1} \right]
\]
\[
= \frac{2\xi}{f(X_{i(k')})} \int \left( \frac{\partial \Psi g}{\partial x_1} \right)(X_{i(k')}, x_2) dx_2,
\]
\[\hfill\]
thus the limit of (10) reads

$$\lim_{K, N \to \infty, \xi \to \xi_0} E_1 = \lim_{K \to \infty} -\mu \xi \sum_{k' = 1}^{K} \left[ \nabla_{x_1} V(X_{i(k')}, X_{j(k')} : \frac{\int (\nabla_{x_1} \Psi g)(X_{i(k')}, x_2)dx_2}{f(X_{i(k')})} \right. $$

$$+ \nabla_{x_1} V(X_{j(k')}, X_{i(k')} : \frac{\int (\nabla_{x_2} \Psi g)(X_{j(k')}, x_2)dx_2}{f(X_{j(k')})} \right]$$

$$= -2\mu \xi \left\langle \left\langle g, \nabla_{x_1} V(x_1, x_2) : \frac{\int (\nabla_{x_1} \Psi g)(x_1, x_2)dx_2}{f(x_1)} \right\rangle \right\rangle.$$

Now, coming back to (9) and performing the same procedure for $E_2$ we obtain

$$\frac{d}{dt} \langle (g(x_1, x_2, t) \Psi(x_1, x_2)) \rangle = -2\mu \xi \left\langle \left\langle g, \nabla_{x_1} V(x_1, x_2) : \frac{\int (\nabla_{x_1} \Psi g)(x_1, x_2)dx_2}{f(x_1)} \right\rangle \right\rangle$$

$$- 2\mu \xi \left\langle \left\langle g, \nabla_{x_2} V(x_1, x_2) : \frac{\int (\nabla_{x_2} \Psi g)(x_2, x_1)dx_2}{f(x_1)} \right\rangle \right\rangle.$$

Integrating by parts, changing the variables and order of integrals we easily obtain

$$\frac{d}{dt} \langle (g(x_1, x_2, t) \Psi(x_1, x_2)) \rangle$$

$$= 2\mu \xi \left\langle \left\langle \nabla_{x_1} : \frac{g(x_1, x_2)}{f(x_1)} \right\rangle \right\rangle \int g \nabla_{x_1} V(x_1, x_2)dx_2 \right\rangle \right\rangle,$$

$$+ 2\mu \xi \left\langle \left\langle \nabla_{x_2} : \frac{g(x_1, x_2)}{f(x_2)} \right\rangle \right\rangle \int g \nabla_{x_2} V(x_2, x_1)dx_1 \right\rangle \right\rangle.$$

Therefore, the differential form of equation for $g$ reads

$$\partial_t g(x_1, x_2, t) = D \left( \Delta_{x_1} g(x_1, x_2, t) + \Delta_{x_2} g(x_1, x_2, t) \right)$$

$$+ 2\mu \xi \nabla_{x_1} : \left( \frac{g(x_1, x_2)}{f(x_1)} \right) F(x_1, t),$$

$$+ 2\mu \xi \nabla_{x_2} : \left( \frac{g(x_1, x_2)}{f(x_2)} \right) F(x_2, t),$$

where we have reintroduced the diffusion terms due to the noise in (3), and $F(x_1)$ is the same one as defined as in (8); recall

$$F(x_1, t) = \int g(x_1, x_2, t) \nabla_{x_1} V(x_1, x_2)dx_2,$$

$$F(x_2, t) = \int g(x_2, x_1, t) \nabla_{x_2} V(x_2, x_1)dx_1.$$

**Step 4.** Equations (8) and (14) do not take into account the phenomena of creation and destruction of links. According to the description at the beginning of this paper, our model describes a process of creation of links with the probability $\nu^N_f$, provided the two particles are sufficiently close to each other. Surely, the number of new links will be proportional to the number of couples of the particles such that one of them is close to $x_1$ and the other one is close to $x_2$, whose distance is less than $R$, this number is equal to

$$\frac{N(N-1)}{2} \int h(x_1, x_2, t) \chi_{|x_1 - x_2| \leq R} dx_1 dx_2 dt,$$
where $h(x_1, x_2, t) = \lim_{N \to \infty} h^N$ and $h^N = h^N(x_1, x_2, t)$ is the two-particle distribution defined in (5). This number has to be decreased by the number of couples that are already connected by existing links:

$$Kg(x_1, x_2, t) \, dx_1 \, dx_2 \, dt.$$ 

Therefore, the number of the new links created during the time interval $[t, t + dt]$ between two points $x_1$ and $x_2$ is equal to

$$\nu^N_f \left( \frac{N(N-1)}{2} h(x_1, x_2, t) \chi_{|x_1 - x_2| \leq R} - Kg(x_1, x_2, t) \right) \, dx_1 \, dx_2 \, dt.$$ 

Dividing this expression by $K$ used for normalization of function $g$ and letting $N, K \to \infty$ so that $K_N \to \xi$ and $\nu^N_f (N - 1) \to \nu_f$ we obtain the probability of creation of the new link equal to

$$\frac{\nu_f}{2 \xi} h(x_1, x_2, t) \chi_{|x_1 - x_2| \leq R}.$$ 

Similarly, the probability that the existing link will be destroyed in the same time interval $[t, t + dt]$ is equal to

$$\nu_d g(x_1, x_2, t),$$

where we used $\nu_d = \lim_{N \to \infty} \nu^N_d$. If we now include these source terms in (14), we get

$$\partial_t g(x_1, x_2, t) = D (\Delta_{x_1} g(x_1, x_2, t) + \Delta_{x_2} g(x_1, x_2, t))$$

$$+ 2\mu \xi \left( \nabla_{x_1} \cdot \left( \frac{g(x_1, x_2)}{f(x_1)} F(x_1, t) \right) + \nabla_{x_2} \cdot \left( \frac{g(x_1, x_2)}{f(x_2)} F(x_2, t) \right) \right)$$

$$+ \frac{\nu_f}{2 \xi} h(x_1, x_2, t) \chi_{|x_1 - x_2| \leq R} - \nu_d g(x_1, x_2, t).$$

This together with (8) gives the system (6). Theorem 2.2 is proved.

Note that system (6) is not closed, since all the three distributions $f$, $g$, and $h$ are a priori unknown. In order to close this system we will have to introduce some closure assumption; this will be done in the next section.

3. Derivation of the macroscopic equations. The equations of distributions of particles and links in the form introduced in Theorem 2.2 do not reveal anything more than relations between certain mechanisms leading to evolution in time of $f$ and $g$. To get somehow deeper insight into the behavior of the system we introduce the characteristic values of the physical quantities appearing in the system. We denote by $t_0$ the unit of time and by $x_0$ the unit of space. A straightforward scaling argument allows us to interpret (8) and (15) obtained in the previous section as scaled equations. Upon choosing time and space units, we can also interpret the coefficients as scaled coefficients in these units. From now on we will use time and space units such that

$$\mu = 1 \quad \text{and} \quad D = 1.$$ 

The next step is to introduce the macroscopic scaling for these units using small parameter $\varepsilon << 1$: $x''_0 = \varepsilon^{-1/2} x_0$, $t''_0 = \varepsilon^{-1} t_0$. Then the new variables and unknowns are

$$\mu = 1 \quad \text{and} \quad D = 1.$$
Our purpose now is to let $\varepsilon$ tend to zero in (16) and (17). Assuming again that $f''$, $g''$, and $h''$ exist we denote $f_\varepsilon = f''$, $g_\varepsilon = g''$, $h_\varepsilon = h''$; we then have the following proposition.

**Proposition 3.1.** Assume that $h_\varepsilon(x_1, x_2) = f_\varepsilon(x_1)f_\varepsilon(x_2)$ and that $V(x, x_j) = U(|x_i - x_j|)$; then provided the limits
\[
\begin{align*}
    f &:= \lim_{\varepsilon \to 0} f_\varepsilon, \\
    g &:= \lim_{\varepsilon \to 0} g_\varepsilon
\end{align*}
\]
exist, they formally satisfy
\[
\begin{align*}
\partial_t f(t, x) &= \Delta_x f(t, x) + \frac{\nu_f}{\nu_d} \nabla_x \cdot (f(t, x) \nabla_x (\tilde{V} \ast f)(t, x)), \\
g(t, x, y) &= \frac{\nu_f}{2\xi \nu_d} f(t, x) f(t, y) \chi_{|x-y| \leq R} 
\end{align*}
\]
(18a)
(18b)
for some compactly supported potential \(\tilde{V}\) specified below.

Proof. Let us start with the limit equation for the distribution of links. From (17), using the assumption on small correlations we obtain
\[
\frac{\nu_f}{2\xi} f_x(t, x) f_x(t, y) \chi_{|x-y| \leq R} - \nu_d g_x(t, x, y) = O(\varepsilon^3).
\]
Letting \(\varepsilon \to 0\) in the above formula, we formally obtain (18b), which is an explicit formula for \(g\). Therefore, plugging this relation into (16) and dropping the tildes again we obtain the equation for \(f\):
\[
\begin{align*}
\partial_t f &= \Delta_x f + \nabla_x \cdot F, \\
F &= \frac{\nu_f}{\nu_d} f(x) \int f(y) \nabla_x V(x, y) \chi_{|x-y| \leq R} dy.
\end{align*}
\]
Taking into account the form of the potential, we can rewrite the above equation in slightly different form,
\[
\partial_t f = \Delta_x f + \frac{\nu_f}{\nu_d} \nabla_x \cdot \left( f(x) \int \nabla \tilde{V}(x-y) f(y) dy \right)
\]
(19)
for some \(\tilde{V}\) such that
\[
\nabla_i \tilde{V}(x) = U'(|x|) \chi_{|x| \leq R \tilde{c}^i}, \quad i = 1, 2,
\]
which gives (18a).

Remark 3.2. The assumption \(h_\varepsilon(x_1, x_2) = f_\varepsilon(x_1) f_\varepsilon(x_2)\) amounts to neglecting spatial correlations for particles; this is a reasonable assumption if each particle interacts with many others. The link distribution described by (18b) looks like that of a random geometric graph, where particles are linked whenever they are distant less than a certain threshold \([25]\); in this case, particles distant less than \(R\) are actually linked only with a certain probability.

Remark 3.3. Equation (19) is an example of an aggregation-diffusion equation with attractive-repulsive potential \(\tilde{V}\). It is difficult to calibrate this equation against any particular phenomena. Let us mention, however, that swarming models often make use of attractive-repulsive potentials such as the one obtained here in the macroscopic description; see, e.g., \([10, 15, 29]\).

4. Analysis of the macroscopic equation: General potential.

4.1. Remark about the free energy. The above system, particularly (18a), is well known in the literature as an aggregation-diffusion equation and as a McKean–Vlasov equation. For analytical and numerical results devoted to solvability and asymptotic analysis of solutions, depending on the shape of the potential \(\tilde{V}\), see, for instance, \([14, 9]\). Concerning the steady states, an exhaustive analysis of this problem would require finding the minima of the following energy functional associated with (18a):
\[
F(f) = \int \left( f \log f + \frac{\nu f}{2 \nu d} f(\hat{V} * f) \right) dx.
\]

It is easy to check that \(F(f)\) is dissipated in time:
\[
\frac{d}{dt} F(f) = \int \left( \Delta f \log f + \frac{\nu f}{\nu d} \Delta f + \frac{\nu f}{\nu d} (\hat{V} * f) \Delta f + \left( \frac{\nu f}{\nu d} \right)^2 \nabla \cdot (f \nabla (\hat{V} * f)) (\hat{V} * f) \right) dx
\]
\[
= -\int \left( \frac{\nabla f}{f} + \left( 2 \frac{\nu f}{\nu d} \right) \nabla (\hat{V} * f) \cdot \nabla f + \left( \frac{\nu f}{\nu d} \right)^2 f |\nabla (\hat{V} * f)|^2 \right) dx
\]
\[
= -\int \left( \frac{\nabla f}{f^{1/2}} + \frac{\nu f}{\nu d} f^{1/2} \nabla (\hat{V} * f) \right)^2 dx \leq 0.
\]

**4.2. Constant steady states.** In this note, we want to focus only on the constant steady states, i.e., \(f_* = \text{const}\), which, on bounded domains, have an interpretation as probability measures. It turns out that the stability or instability of the steady states for (18a) is related to the notion of H-stability of the potential \(\hat{V}\). According to the definitions from classical statistical mechanics, the compactly supported potential \(\hat{V}\) is H-stable provided the integral \(\int_{\mathbb{R}^2} \hat{V}(x) dx\) is positive; otherwise it is not H-stable (unstable) [27]. For the H-stable potentials, the aggregation part of (18a) acts as diffusion, so, any initial perturbation is smoothed infinitely fast. For potentials that are not H-stable, the asymptotical behavior of the solution is much more interesting. For our system in its general form we only prove the following criterion for instability of the constant steady states.

**Lemma 4.1.** Let the potential \(\hat{V}\) be integrable and let
\[
M = \int_{\mathbb{R}^2} \hat{V}(x) dx < 0.
\]
Then the constant steady state \(f_*\) is unstable if
\[
f_* > -\frac{1}{M} \frac{\nu d}{\nu f}.
\]

**Proof.** In order to check the stability of the constant steady state \(f_* > 0\), we linearize (18a) around \(f_*\). We assume that \(f\) is a small perturbation of \(f_* \) (\(f << f_*\)) and thus \(f\) satisfies
\[
\partial_t f(t, x) = \Delta_x f(t, x) + f_* \frac{\nu f}{\nu d} \Delta_x ((\hat{V} * f)(t, x)).
\]
Then we apply the Fourier transform in space to both sides of (22); we obtain
\[
\partial_t \hat{f}(t, y) = -y^2 \hat{f}(t, y) - 2\pi f_* \frac{\nu f}{\nu d} y^2 \hat{V}(t, y).
\]
The Taylor expansion around zero of the Fourier transform of \(\hat{V}\) is equal to
\[
\hat{V}(y) = \frac{1}{2\pi} \int_{\mathbb{R}^2} \hat{V}(x) dx + O(y) = \frac{M}{2\pi} + O(y).
\]
Plugging it into (23) we obtain

\[ \partial_t \log \hat{f}(t, y) = - \left(1 + \int \frac{\nu f}{\nu d} M \right) y^2 + O(y^3), \]

and so, for negative \( M \), we can always find sufficiently large \( f_* \) leading to instability of the steady state \( f_* \). More precisely, for (21) the right-hand side (r.h.s.) of (24) for sufficiently small \( y \) is larger than some positive constant \( c \); thus

\[ \hat{f}(t) \geq \hat{f}_0 e^{ct} \rightarrow \infty \quad \text{for} \quad t \rightarrow \infty, \]

and so the steady state \( f_* \) is unstable. \( \square \)

5. Analysis of the macroscopic equation: Hookean potential.

5.1. Preliminaries. Until this moment, the exact form of potential (1) did not play any role and we could work assuming only its symmetricity and integrability. Let us now focus on a particular form. If we imagine that the links between the particles play any role and we could work assuming only its symmetricity and integrability. Let us now compute the integral of our potential \( \tilde{V} \). First, it is easy to see that \( \tilde{V} \) denotes the rest length of the spring and the intensity parameter \( \kappa \) is a positive number, characteristic of the spring. We then have

\[ \int f(y) \chi_{|y| < R} \nabla x V(x, y) dy = \int f(y) \kappa (|x - y| - l_0) \frac{x - y}{|x - y|} \chi_{|x - y| < R} dy. \]

We now want to find \( \tilde{V} \) such that the equation for \( f \) is in the form (19). In our case \( \tilde{V}(x) \) satisfies \( \nabla_i \tilde{V}(x) = \kappa (|x| - l_0) \chi_{|x| \leq R} \), where \( x \in \mathbb{R}^2 \), moreover \( \tilde{V}(x) = 0 \) for \( |x| > R \). First, it is easy to see that \( \tilde{V}(x) \) is a radially symmetric function, and thus we can introduce \( U(|x|) = \tilde{V}(x) \); second, since the potential \( U(r) \) vanishes for \( r \geq R \) we have

\[ U(2R) - U(r) = \int_r^R (s - l_0) ds = \frac{\kappa}{2} \left( (R - l_0)^2 - (r - l_0)^2 \right). \]

Therefore, \( U(r) = \frac{\kappa}{2} \left( (r - l_0)^2 - (R - l_0)^2 \right) \), and so

\[ \tilde{V}(x) = \begin{cases} \frac{\kappa}{2} \left( |x| - l_0 \right)^2 - (R - l_0)^2 & \text{for} \quad |x| < R, \\ 0 & \text{for} \quad |x| \geq R; \end{cases} \]

see Figure 1.

Let us now compute the integral of our potential \( \tilde{V} \) given in (25). We have

\[ \int_{\mathbb{R}^2} \tilde{V}(x) dx = \frac{\kappa}{2} \int_{|x|<R} \left( (|x| - l_0)^2 - (R - l_0)^2 \right) \chi_{|x|<R} dx \]

\[ = \pi \kappa \int_0^R \left( (r - l_0)^2 - (R - l_0)^2 \right) r dr \]

\[ = \pi \kappa \left( \frac{r^4}{4} - \frac{2r^3l_0}{3} - R^2 r^2 + R l_0 r^2 \right) \bigg|_0^R \]

\[ = \pi \kappa R^3 \left( \frac{l_0}{3} - \frac{R}{4} \right); \]
therefore, according to the definition given above, \( \tilde{V} \) is H-stable if the condition \( l_0 > \frac{3R}{4} \) is satisfied. Lemma 4.1 provided a special criterion for the constant steady state to be unstable, and this is basically all the information we can get for the whole space case. However, if we now consider the same problem on the space periodic domain the criteria obtained in Lemma 4.1 will have to include the size of the domain. Moreover, it can happen that even if unstable, the steady state might be only weakly unstable, meaning that only one mode from a countable set of modes will be unstable, while the rest of them will be stable. The intention of the linear analysis in the whole space case presented below is to provide some intuition on the behavior of the potential, so that it is more intuitive how to “select” the unstable modes in the second part of this section.

### 5.2. Linear analysis in the whole space

To understand the behavior of the solutions close to the stability/instability threshold (21) we come back to (23) and we compute the Fourier transform of \( \tilde{V} \) given by (25),

\[
\mathcal{F} \tilde{V}(y) = \frac{1}{2\pi} \int_{\mathbb{R}^2} e^{-ix \cdot y} \tilde{V}(x) \, dx.
\]

Due to the radial symmetry of \( \tilde{V} \), our transform gives radially symmetric function \( \mathcal{F} \tilde{V}(s) = \tilde{V}(s) \), where \( s = |y| \), that satisfies

\[
\mathcal{F} \tilde{V}(s) = \frac{1}{2\pi} \int_{0}^{2\pi} \int_{0}^{\infty} e^{-i \theta \cos(\theta)} \tilde{V}(r) r \, dr \, d\theta.
\]

\[
= \frac{\kappa}{2} \int_{0}^{sR} \left[ \left( \frac{h}{s} - l_0 \right)^2 - (R-l_0)^2 \right] J_0(h) \frac{h}{s^2} \, dh
\]

\[
= \frac{\kappa(2l_0-R)R}{2s^2} \int_{0}^{sR} h J_0(h) \, dh - \frac{\kappa l_0}{s^3} \int_{0}^{sR} h^2 J_0(h) \, dh + \frac{\kappa}{2s^4} \int_{0}^{sR} h^3 J_0(h) \, dh,
\]

where \( J_0 \) is the Bessel function of the first kind of order 0. In order to compute integrals of the type \( \int_{0}^{sR} h^\alpha J_0(h) \, dh \) for \( \alpha = 1, 2, 3 \), we recall the Maclaurin series for
the Bessel function of order $i$,
\[ J_i(x) = \sum_{m=0}^{\infty} \frac{(-1)^m}{m!i^{m+i+1}} \left(\frac{x}{2}\right)^{2m+i}, \]
and for the Struve functions of order $i$,
\[ H_i(x) = \sum_{m=0}^{\infty} \frac{(-1)^m}{\Gamma(m + 3/2)\Gamma(m + i + 3/2)} \left(\frac{x}{2}\right)^{2m+i+1}. \]

Using this notation (26) gives
\[ \hat{V}(s) = \kappa \left( J_0(sR) \frac{R^2}{s^2} - J_1(sR) \frac{2R}{s} + \frac{\pi RL_0}{2s^2} [J_1(sR)H_0(sR) - J_0(sR)H_1(sR)] \right). \]

Therefore, the general equation (23) now has the following form:
\[ \partial_t \log f(t,y) = -y^2 - 2\pi f \sum_{d}^{\nu_d} \left( J_0(|y|)R^2 - J_1(|y|) \frac{2R}{|y|} \right) \]
\[ + \frac{\pi RL_0}{2} [J_1(|y|)H_0(|y|)R - J_0(|y|)H_1(|y|)]. \]

We now write an explicit form of the solution emanating from the initial data $f(0) = f_0$,
\[ f(t,y) = f_0(y)e^{-G(y)t}, \]
where the exponent $G = G(y, R, l_0, \kappa, \nu_f, \nu_d, f_0)$ is given by
\[ G = y^2 + 2\pi f \sum_{d}^{\nu_d} \left( J_0(|y|)R^2 - J_1(|y|) \frac{2R}{|y|} \right) \]
\[ + \frac{\pi RL_0}{2} [J_1(|y|)H_0(|y|)R - J_0(|y|)H_1(|y|)]. \]

From Lemma 4.1 we know exactly when $G$ ceases to be nonnegative close to $y = 0$. Let us now see what happens slightly further from the origin. To this purpose, we rewrite (27) in the following form:
\[ G(z)R^2 = z^2 + 2\pi f \sum_{d}^{\nu_d} R^4 \left( \frac{\pi l_0}{2R^2} [J_1(z)H_0(z) - J_0(z)H_1(z)] - J_2(z) \right), \]
where $z = |y| R$. To investigate the minima of $G(z)$ we check the minima of another function, namely,
\[ F^{\alpha,\beta}(z) = G(z)R^2 = z^2 + \beta \left( \frac{\pi \alpha}{2} [J_1(z)H_0(z) - J_0(z)H_1(z)] - J_2(z) \right), \]
where the parameters $\alpha, \beta > 0$ are related to $R, l_0, \kappa, \nu_f, \nu_d, f_0$ in the following way:
\[ \alpha = \frac{l_0}{R}, \quad \beta = \frac{2\pi \kappa f \nu_f R^4}{\nu_d}. \]

The interesting range for parameter $\alpha$ is $[0,1]$ and for the parameter $\beta$ we take $[0, \infty)$. In Figure 2, we present the graphs of the two functions $\frac{\pi}{2} [J_1(z)H_0(z) - J_0(z)H_1(z)]$ and $-J_2(z)$ that are included in the definition of $F^{\alpha,\beta}(z)$ from (28). Note that from (28) it is clear that $F^{\alpha,\beta}(0) = 0$ for all values of $\alpha, \beta$. On the other hand, the picture above suggests that changing the values of parameters $\alpha, \beta$ may cause that $F^{\alpha,\beta}$ will achieve negative values. In particular, by choosing a sufficiently small value for parameter $\alpha$ we would get a negative value of $\frac{\pi \alpha}{2} [J_1H_0 - J_0H_1] - J_2$ close to zero. This is nothing other than rephrasing the criterion from Lemma 4.1 in terms of $\alpha$ and $\beta$. 

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Fig. 2. The graph of functions $\frac{\pi}{2}[J_1(z)H_0(z) - J_0(z)H_1(z)]$ and $-J_2(z)$. Decreasing the value of parameter $\alpha$ in (28) decreases the amplitude of the oscillations marked with the continuous line which leads to negative value of $F^{\alpha,\beta}(z)$ close to $z = 0$.

**Proposition 5.1.** Let $\alpha$ and $\beta$ be given as in (29); then if $(\alpha, \beta) \in U_{\mathbb{R}^2}$, where

$$U_{\mathbb{R}^2} = \left\{ (\alpha, \beta) \in [0, 1] \times [0, \infty) : \alpha < \frac{3}{4}, \beta > \frac{24}{3 - 4\alpha} \right\},$$

the steady state $f_*$ is unstable, and otherwise it is stable.

**Proof.** Instability of the steady state follows as previously from expansion of $F^{\alpha,\beta}(z)$ in the neighborhood of $z = 0$. After lengthy but straightforward calculations we obtain

$$F^{\alpha,\beta}(z) = \left( 4 + \beta \frac{2\alpha}{3} - \beta \frac{1}{2} \right) \left( \frac{z}{2} \right)^2 + O(z^4).$$

Finally, we see that taking $\alpha < \frac{3}{4}$ we can always find sufficiently large $\beta$ (i.e., $\beta > \frac{24}{3 - 4\alpha}$), so that the first term is negative and hence, for small enough $z$, the whole $F^{\alpha,\beta}(z)$ is negative as well. The fact that for parameters $(\alpha, \beta) \not\in U_{\mathbb{R}^2}$, the steady state is stable is shown numerically. In Figure 3, we present the minimum of $F^{\alpha,\beta}$ with respect to $z$, i.e.,

$$F_{\min}^{\alpha,\beta} = \min_{z \in [0, 1]} F^{\alpha,\beta}(z),$$

as a function of parameters $\alpha, \beta$. The flat region corresponds to the parameter configuration that causes the minimum of $F^{\alpha,\beta}(z)$ to be attained at $z = 0$ and is equal to 0.

**Remark 5.2.** Proposition 5.1 provides some qualitative insight into the behavior of the physical or biological systems that could be modeled by such dynamical networks. The connectivity of the network appears through the parameter $\nu_f/\nu_d$: instability of homogeneous states is favored by a strongly connected network. The $\alpha$ parameter measures the relative importance of the repulsive and attractive parts of the interaction; a more attractive interaction favors instability.

**Remark 5.3.** Once $f$ is known, (18b) allows us to make a connection with the distribution of links of the underlying network. It follows, in particular, that in the
case when distribution of the particles is homogeneous, so is the distribution of the links. Moreover, when the density distribution develops spatial inhomogeneities, so does the link distribution.

5.3. Linear analysis in the spatially periodic case. Let us now investigate the same equation (22) but in the case of the space periodic domain. We will check an influence of the size of the domain on the stability of stationary solutions. The analysis of what happens with the solution in the unstable regime, but close to the instability threshold, will be presented in the next section.

We start by expanding our solution $f(x)$ for $x = (x_1, x_2) \in [-L_1, L_1] \times [-L_2, L_2]$ into the Fourier series. Introducing the shorthand notation for the Fourier modes

$$e_{k_1,k_2} = \exp \left[ i \pi \left( \frac{k_1 x_1}{L_1} + \frac{k_2 x_2}{L_2} \right) \right],$$

we may write

$$f(x_1, x_2) = \sum_{k_1,k_2 \in \mathbb{Z}} \hat{f}_{k_1,k_2} e_{k_1,k_2},$$

where the Fourier coefficients $\hat{f}_{k_1,k_2}$ are given by

$$\hat{f}_{k_1,k_2} = \frac{1}{4L_1L_2} \int_{-L_2}^{L_2} \int_{-L_1}^{L_1} f(x_1, x_2) e^{-k_1 x_1 - k_2 x_2} \, dx_1 \, dx_2.$$  

Recall that we have the properties for the Fourier coefficients of the derivatives of functions

$$\partial_{x_1}^n f_{k_1,k_2} = \left( -i \frac{\pi k_1}{L_1} \right)^n \hat{f}_{k_1,k_2}, \quad \partial_{x_2}^n f_{k_1,k_2} = \left( -i \frac{\pi k_2}{L_2} \right)^n \hat{f}_{k_1,k_2}.$$
and of the convolution of functions
\[ \hat{f} \ast g_{k_1, k_2} = \left[ \int_{-L_2}^{L_2} \int_{-L_1}^{L_1} f(x-y)g(y) \, dy \right]_{k_1, k_2} = 4L_1L_2 \hat{f}_{k_1, k_2} \hat{g}_{k_1, k_2}. \]

Therefore, multiplying both sides of linearized system (22) by \( \frac{1}{4L_1L_2} e^{-k_1 \cdot -k_2} \) and integrating over \([-L_1, L_1] \times [-L_2, L_2]\), we obtain

\[
\partial_t \hat{f}_{k_1, k_2} = -\pi^2 \left( \frac{k_1^2}{L_1^2} + \frac{k_2^2}{L_2^2} \right) \hat{f}_{k_1, k_2} - f_* \frac{v_f}{v_d} \pi^2 \left( \frac{k_1^2}{L_1^2} + \frac{k_2^2}{L_2^2} \right) 4L_1L_2 \hat{V}_{k_1, k_2} \hat{f}_{k_1, k_2}.
\]

This time \( f_* \) can be interpreted as a probability measure; thus from now on, we will take \( f_* = \frac{1}{\pi R^d L_1^d} \), which on the rectangle \([-L_1, L_1] \times [-L_2, L_2]\) integrates to one, and so, for any \( k_1, k_2 \in \mathbb{Z} \), we obtain

\[
\hat{f}_{k_1, k_2}(t) = \hat{f}_0(k_1, k_2) e^{-G_{k_1, k_2}t},
\]

where
\[
G_{k_1, k_2} = \pi^2 \left( \frac{k_1^2}{L_1^2} + \frac{k_2^2}{L_2^2} \right) + \frac{v_f}{v_d} \pi^2 \left( \frac{k_1^2}{L_1^2} + \frac{k_2^2}{L_2^2} \right) \hat{V}_{k_1, k_2}.
\]

To compute \( \hat{V}_{k_1, k_2} \) in the case when \( R < \min\{L_1, L_2\} \) we write
\[
\hat{V}_{k_1, k_2} = \frac{1}{4L_1L_2} \int_0^{2\pi} \int_0^R e^{-i\pi \sqrt{\frac{k_1^2}{L_1^2} + \frac{k_2^2}{L_2^2}}} \hat{V}(r) r \, dr \, d\theta
\]
\[
= \frac{\pi}{2L_1L_2} \int_0^R \hat{V}(r) J_0 \left( \pi \sqrt{\frac{k_1^2}{L_1^2} + \frac{k_2^2}{L_2^2}} \right) r \, dr
\]
and the last integral can be computed exactly as in the previous section so that we get
\[
\hat{V}_{k_1, k_2} = \frac{\pi R^d L_1}{2L_1L_2} \left( \frac{\pi R^d L_1}{2\pi R^d L_1} \right) \left( J_1(z_{k_1, k_2}) H_0(z_{k_1, k_2}) - J_0(z_{k_1, k_2}) H_1(z_{k_1, k_2}) \right)
\]
\[
- J_2(z_{k_1, k_2}) \frac{R^4}{2z_{k_1, k_2}},
\]
where we denoted
\[
z_{k_1, k_2} = \pi R \sqrt{\frac{k_1^2}{L_1^2} + \frac{k_2^2}{L_2^2}},
\]
and so
\[
F^{\alpha, \beta}(z_{k_1, k_2}) = G_{k_1, k_2} R^2 = z_{k_1, k_2}^2
\]
\[
+ \beta \left( \frac{\pi \alpha}{2} \left( J_1(z_{k_1, k_2}) H_0(z_{k_1, k_2}) - J_0(z_{k_1, k_2}) H_1(z_{k_1, k_2}) \right) - J_2(z_{k_1, k_2}) \right),
\]
for parameters \( \alpha \) and \( \beta \) such that
\[
\alpha = \frac{l_0}{R}, \quad \beta = \frac{\pi v_f R^4}{2v_d L_1 L_2}.
\]
Note that these are the same parameters as in (29) with $f_\star = \frac{1}{4L_1L_2}$. Moreover, function $F_{\alpha,\beta}$ has the same form as in the whole space case (28) but is evaluated only at the discrete set of points $z_{k_1,k_2}$, $k_1, k_2 \in \mathbb{Z}$. We know already that for continuous arguments $z \in [0, \infty)$ there is a phase transition curve $\beta(\alpha) = \frac{24}{32-4\alpha}$. The proof of this fact was based on finding a negative value of $F_{\alpha,\beta}(z)$ sufficiently close to $z = 0$. Here, however, the discrete variable $z_{k_1,k_2}$ depends on the size of the domain and it may happen that $F_{\alpha,\beta}(z_{k_1,k_2})$ for all $k_1, k_2 \in \mathbb{Z}$ is always positive even if $F_{\alpha,\beta}(z)$ does attain negative value. Indeed, we have the following proposition.

**Proposition 5.4.** For a nonempty subset of parameters $(\alpha, \beta) \in U_{\mathbb{R}^2}$, there exist $L_1, L_2 \in [R, \infty)$ such that $f_\star = \frac{1}{4L_1L_2}$ is a stable solution of (18a).

**Proof.** The proof of this fact is again numerical. Figure 4 illustrates the function $F_{\alpha,\beta}(z)$ in the unstable range of $\alpha, \beta$. Note that if $z_{1,0}, z_{0,1}$ are larger than $z_0 = 0.63$ the steady state $f_\star$ will not be affected by the unsteady modes. In Figure 5 we present

![Figure 4](image1)

**Fig. 4.** Left: the function $F_{\alpha,\beta}(z)$ for $\alpha = 0.5, \beta = 25$; right: the zoom of the graph in the neighborhood of $z = 0$; $z_{\min}$ denotes the point where the minimum of $F_{\alpha,\beta}(z)$ is attained, while $z_0$ denotes the first zero of the function $F_{\alpha,\beta}(z)$ for $z > 0$.

![Figure 5](image2)

**Fig. 5.** Left: the positions of minima of function $F_{\alpha,\beta}(z)$, $z_{\min}(\alpha, \beta)$; right: the positions of zero of $F_{\alpha,\beta}(z)$, $z_0(\alpha, \beta)$ for variable parameters $\alpha \in [0, 1]$ and $\beta \in [0, 40]$. 

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the positions of minima of function $F^{\alpha,\beta}(z)$, $z_{\min}(\alpha, \beta)$ and the positions of zero of $F^{\alpha,\beta}(z)$, $z_0(\alpha, \beta)$. We see in particular that $z_0(\alpha, \cdot)$ is a monotonically increasing function, while $z_0(\cdot, \beta)$ is monotonically decreasing. However, from (34) we get

$$z_{1,0} = \frac{\pi R}{L_1} \leq \pi \quad \text{and} \quad z_{0,1} = \frac{\pi R}{L_2} \leq \pi;$$

therefore, the statement can be fulfilled, for example, for $L_1 = L_2 = R$ and $\alpha^*, \beta^*$ such that

$$z_0(\alpha^*, \beta^*) < \pi; \quad (35)$$

since $|z_0(\alpha, \beta)| \geq |z_{\min}(\alpha, \beta)|$, the pair of parameters $(\alpha^*, \beta^*) \in U_{R^2}$.

The condition (35) can be rephrased as

$$F^{\alpha,\beta}(\alpha^*, \beta^*)(\pi) > 0,$$

which gives $\beta^*(0.7332\alpha^* - 0.4854) > -9.8696$. This means in particular that $\alpha^* \in (0.5499, 0.75)$ and any $\beta^* \in [0, \infty)$ the stationary solution $f_* = \frac{1}{4\pi^2}$ is a stable solution to (18a) on a periodic box $[-R, R]^2$.

Using the same argument, we can also show the reverse statement to Proposition 5.4.

**Proposition 5.5.** For every $L_1, L_2 \in [R, \infty)$, there exists a nonempty subset of parameters $(\alpha, \beta) \in U_{R^2}$ such that $f_* = \frac{1}{4L_1L_2}$ is a stable solution of (18a).


6.1. Preliminaries. The purpose of this section is to investigate the qualitative behavior of the model beyond the linear level. We will choose the parameters $\alpha, \beta$ in the unstable regime, but close to the stability/instability threshold. In particular, the instability will be associated only with the first nontrivial modes, and the instability rate will be assumed small. As we saw in the previous section this can be guaranteed by the appropriate choice of the size of periodic domain.

The analysis will be made for periodic domains of two types: the rectangular periodic domain and the square periodic domain. As we will see below, in the case when one side of the periodic domain is larger than the other, we may select only one unstable mode and reduce the analysis to a one-dimensional problem. For the case of a square box, the extra symmetry induces a degeneracy of the unstable mode. In both cases we give precise conditions for continuous and discontinuous phase transitions. In the end of this section we also provide numerical verification of these conditions for the Hooke potential. Computations with other domains are in principle possible but would be more complicated and/or less explicit. Nevertheless, we expect that in absence of special symmetries, the picture for a generic domain would look like the one for a periodic rectangle.

Our analysis allows us to identify two types of steady states for the density distribution in the macroscopic model: the homogeneous steady state $f_*$ and the inhomogeneous steady states in the unstable regime. Concerning the network, it is never constant since links are created and destroyed continuously, but the regions of high particle density are regions of high connectivity. It follows from (18b) that if $f$ settles into a stationary state, the network becomes stationary in a probabilistic sense. If a homogeneous $f_*$ is unstable, then the network also develops spatial inhomogeneities; see Remark 5.3.
We would like to emphasize that the theoretical results presented in this section are applicable to a much wider class of potentials under mild assumptions on the Fourier coefficients as stated in Theorems 6.1 and 6.3 for the rectangular and the square case, respectively. The Hookean potential should be treated only as an example for which the more explicit computations and numerical verification are possible. Our starting point is (18a), which we recall here for convenience:

\[ \partial_t f = \Delta f + \gamma \nabla \cdot (f \nabla (\hat{V} \ast f)) \]

with \( \gamma = \frac{\nu}{\nu_c} \).

6.2. The rectangular case for general potential—nondegenerate. We start our analysis from the simpler case when the periodic domain is rectangular, 

\[ (x_1, x_2) \in [-L_1, L_1] \times [-L_2, L_2] \quad \text{such that} \quad L_1 > L_2 \]

and that only the modes \((\pm 1, 0)\) are unstable; all the others are stable. Having in mind the argument from the previous section, this is possible for some \((\alpha^*, \beta^*) \in \mathbb{R}^2\) provided

\[ z_{1,0} < z_0(\alpha^*, \beta^*) < z_{2,0} \quad \text{and} \quad z_{0,1} > z_0(\alpha^*, \beta^*). \]

Looking at the problem from the perspective of stable and unstable modes, we see that an analogous condition can be deduced directly from (32). Namely, the eigenvalue associated with the first mode in the direction \(x_1\) should be the only positive one. This results in the conditions

\[ \lambda = \lambda_{\pm 1,0} = -\frac{\pi^2}{L_1^2} \left( 1 + \gamma \hat{V}_{1,0} \right) > 0, \]

\[ \lambda_{k_1, k_2} = -\pi^2 \left( \frac{k_1^2}{L_1^2} + \frac{k_2^2}{L_2^2} \right) \left( 1 + \gamma \hat{V}_{k_1, k_2} \right) < 0 \quad \text{for} \quad (k_1, k_2) \neq (\pm 1, 0). \]

Recalling notation (31), the unstable modes are then

\[ e_{1,0} = e^{\frac{i \pi x_1}{L_1}} \quad \text{and} \quad e_{-1,0} = e^{\frac{-i \pi x_1}{L_1}}. \]

We now want to check what happens with the constant steady state after passing the instability threshold. We could, for example, think of fixing the parameter \(\alpha\) according to Proposition 5.5 and slowly increase parameter \(\beta\) by changing the value of \(R\). Alternatively, one can identify the instability threshold with changing the sign of \(\lambda\)—this is the standard strategy in bifurcation theory and the one we follow here.

After crossing the instability threshold, one expects that the solution to the nonlinear problem behaves for a short time like the linearized solution, that is, an exponential in time times the unstable mode:

\[ f = f_* + A(t)e_{1,0} + A^*(t)e_{1,0} \quad \text{with} \quad A(t) \propto e^{\lambda t}. \]

Then, if \(A(t)\) remains small, one can hope to expand the solution into power series of \(A(t)\),

\[ f = f_* + A(t)e_{1,0} + A^*(t)e_{1,0} + O(A(t)^2). \]

The goal is then to find a reduced equation for \(A(t)\) that would allow us to understand the dynamics of the solution just by analyzing an ODE for \(A(t)\) (central manifold reduction). The unstable eigenvalue is real, and the system is translation-symmetric. Hence we expect a pitchfork bifurcation when \(\lambda\) changes sign from “−” to “+,” with two possible scenarios:
As mentioned above, at a linear order, $A(t)$ first grows exponentially, but then $f$ tends to an almost homogeneous stationary state:

- a supercritical bifurcation: $A(t)$ grows exponentially, and $f$ tends to an almost homogeneous stationary state.

- a subcritical bifurcation: $A(t)$ grows exponentially until it leaves the perturbative regime, then the final state may be very far from the original homogeneous state.

Instead of adopting a dynamical approach as done here, bifurcations for systems such as (36) can be studied from a “thermodynamical” point of view, i.e., by looking at the minimizers of (20). This has been done in particular in [14]. The second order phase transition in [14] corresponds to the supercritical scenario described above, while the first order phase transition corresponds to the subcritical scenario. However, one should note that the dynamical bifurcation point (where $\lambda$ changes sign) does not coincide with the first order phase transition parameters; the dynamical bifurcation would rather be called a spinodal point in thermodynamics, the language of [14].

The main result of this section provides a criterion allowing us to distinguish these two cases.

**Theorem 6.1.** Assume that $\lambda > 0$ and that $\lambda k_1 k_2 < 0$ for any $(k_1, k_2) \neq (\pm 1, 0)$. Then, there are two possibilities:

- for $2\tilde{V}_{2,0} - \tilde{\nu}_{-1,0} > 0$ the steady state exhibits a supercritical bifurcation,
- for $2\tilde{V}_{2,0} - \tilde{\nu}_{-1,0} < 0$ the steady state exhibits a subcritical bifurcation.

**Proof.** We now want to investigate the evolution of the perturbation $g$ of the constant steady state $f_*$. Hence, the solution to (18a) has the form $f = f_* + \eta$. We denote the operator associated with the linearized equation (22) by $\mathcal{L}(f)$; more precisely,

$$
\partial_t \eta(t, x) = \Delta_x \eta(t, x) + \gamma f_* \Delta_x ((\tilde{V} \ast \eta)(t, x)) := \mathcal{L}(\eta).
$$

Note that $\mathcal{L}(\eta)$ with periodic boundary conditions is a self-adjoint operator. Next, we also distinguish the nonlinear part of (18a) and we denote it by $\mathcal{N}(\eta)$; this gives

$$
(38) \quad \partial_t \eta = \mathcal{L}(\eta) + \mathcal{N}(\eta),
$$

where

$$
\mathcal{N}(\eta) = \mathcal{Q}(\eta, \eta), \quad \mathcal{Q}(\eta_1, \eta_2) = \gamma \nabla \cdot \left( \eta_1 \nabla (\tilde{V} \ast \eta_2) \right)
$$

with $\gamma = \frac{\nu^2}{\kappa_d}$. In what follows we will need to compute the action of $\mathcal{L}$ and $\mathcal{Q}$ on the Fourier basis. We have

$$
(39) \quad \mathcal{L}(e_{k_1, k_2}) = \left[ -\pi^2 \left( \frac{k_1^2}{L_1^2} + \frac{k_2^2}{L_2^2} \right) (1 + \gamma \tilde{V}_{k_1, k_2}) \right] e_{k_1, k_2} = \lambda_{k_1, k_2} e_{k_1, k_2},
$$

$$
(40) \quad \mathcal{Q}(e_{k_1, k_2}, e_{l_1, l_2}) = -4L_1 L_2 \gamma \pi^2 \tilde{V}_{l_1, l_2} \left( \frac{l_1(k_1 + l_1)}{L_1^2} + \frac{l_2(k_2 + l_2)}{L_2^2} \right) e_{k_1 + l_1, k_2 + l_2}
$$

As mentioned above, at a linear order, $\eta$ moves on a vector space spanned by $e_{1,0}, e_{-1,0}$,

$$
\eta(t, x) = A(t)e_{1,0} + A^*(t)e_{-1,0}.
$$

Furthermore, if the equation were linear, the solution emanating from any initial condition would be quickly attracted toward this vector space. This follows from the fact that all the other modes of motion are stable. For the nonlinear system, we
expect that \( \text{span}(e_{1,0}, e_{-1,0}) \) will be deformed into some manifold. This manifold is tangent to \( \text{span}(e_{1,0}, e_{-1,0}) \) close to \( \eta = 0 \) and can be parametrized by the projection of \( \eta \) on this space,

\[
\eta(t, x) = A(t)e_{1,0} + A^*(t)e_{-1,0} + H[A, A^*](x),
\]

with \( H \) such that

\[
H[A, A^*] = O(A^2, AA^*, (A^*)^2) \quad \text{and} \quad \langle e_{1,0}, H \rangle = \langle e_{-1,0}, H \rangle = 0.
\]

Furthermore, from translation invariance we can write, using Lemma 6.2 (see below),

\[
H[A, A^*] = \sum_{k_1 \geq 0} A^{k_1} h_{k_1,0}(\sigma)e_{k_1,0} + \sum_{k_1 < 0} (A^*)^{-k_1} h_{k_1,0}(\sigma)e_{k_1,0},
\]

where

\[
\sigma = |A|^2 \quad \text{and} \quad h_{k,0} = h_{k,0}^0 + \sigma h_{k,0}^1 + \cdots.
\]

The conditions (42) imply that \( h_{1,0} = h_{-1,0} = 0 \). Moreover, \( h_{k_1,0}^0 = 0 \) for \( k_1 = 0, \pm 1 \); otherwise \( H[A, A^*] \) would contain zero and first order terms in \( A, A^* \). Hence, at the leading order, only the modes \( (\pm 2, 0) \) remain; more precisely

\[
H[A, A^*] = A^2 h_{2,0}^0 e_{2,0} + (A^*)^2 h_{-2,0}^0 e_{-2,0} + O((A, A^*)^3).
\]

Then, plugging (41) and (43) into the definitions of \( \mathcal{L}(\eta) \) and \( \mathcal{N}(\eta) \) we obtain

\[
\mathcal{L}(\eta) = A\mathcal{L}(e_{1,0}) + A^*\mathcal{L}(e_{-1,0})
\]

\[
+ A^2 h_{2,0}^0 \mathcal{L}(e_{2,0}) + (A^*)^2 h_{-2,0}^0 \mathcal{L}(e_{-2,0}) + O((A, A^*)^3)
\]

and

\[
\mathcal{N}(\eta) = A^2 \mathcal{Q}(e_{1,0}, e_{1,0}) + (A^*)^2 \mathcal{Q}(e_{-1,0}, e_{-1,0})
\]

\[
+ A^3 h_{2,0}^0 [\mathcal{Q}(e_{1,0}, e_{2,0}) + \mathcal{Q}(e_{2,0}, e_{1,0})]
\]

\[
+ |A|^2 A h_{2,0}^0 [\mathcal{Q}(e_{-1,0}, e_{2,0}) + \mathcal{Q}(e_{2,0}, e_{-1,0})]
\]

\[
+ (A^*)^3 h_{-2,0}^0 [\mathcal{Q}(e_{-1,0}, e_{-2,0}) + \mathcal{Q}(e_{-2,0}, e_{-1,0})]
\]

\[
+ |A|^2 A^* h_{-2,0}^0 [\mathcal{Q}(e_{1,0}, e_{-2,0}) + \mathcal{Q}(e_{-2,0}, e_{1,0})]
\]

\[
+ O((A, A^*)^4).
\]

Therefore, the full dynamics of \( \eta \) can be obtained by substituting the above formulas for \( \mathcal{L}(\eta) \) and \( \mathcal{N}(\eta) \) into (38). On the other hand, differentiating (41) with respect to time, and using (43), we have

\[
\partial_t \eta = \dot{A} e_{1,0} + \dot{A}^* e_{-1,0} + 2A \dot{A} h_{0,2,0}^0 e_{2,0} + 2A^* \dot{A}^* h_{0,2,0}^0 e_{-2,0} + \partial_t O((A, A^*)^3).
\]

We now equate expressions \( \partial_t \eta = (44) + (45) \) and (46) and compare Fourier mode by Fourier mode, and order in \( A \) by order in \( A \). We start with the mode \( e_{1,0} \). Taking the scalar product of the r.h.s. of (44) and (45) with \( e_{1,0} \), we get

\[
\langle e_{1,0}, \mathcal{L}(\eta) \rangle = A(e_{1,0}, \mathcal{L}(e_{1,0}))
\]
and
\[ \langle e_{1,0}, \mathcal{N}(\eta) \rangle = |A|^2 A h_{2,0}^0 (e_{1,0}, Q(e_{-1,0}, e_{2,0}) + Q(e_{2,0}, e_{-1,0})) + O((A, A^*)^4), \]
where \( \langle u, v \rangle = 1/(4L_1 L_2) \int_{-L_2}^{L_2} \int_{-L_1}^{L_1} u^* v \, dx_1 \, dx_2 \). Comparing these expressions with the projection of (46) on \( e_{1,0} \) we obtain
\[ \dot{A}(e_{1,0}, e_{1,0}) = A(e_{1,0}, L(e_{1,0})) + |A|^2 A h_{2,0}^0 (e_{1,0}, Q(e_{-1,0}, e_{2,0}) + Q(e_{2,0}, e_{-1,0})) + O((A, A^*)^4). \]
So, using (39) and (40) we obtain
\[ \dot{A} = A \lambda + |A|^2 A h_{2,0}^0 \gamma \pi^2 4 L_2 \frac{L_1}{L} \left( \hat{V}_{-1,0} - 2 \hat{V}_{2,0} \right) + O((A, A^*)^4). \]
The terms of the leading order in \( A \) yield the linearized dynamics. To investigate the behavior of \( A \) at the nonlinear level we need first to compute \( h_{2,0}^0 \): we do this by equating the Fourier coefficient (2, 0) in \( \partial_t \eta = (44) + (45) \) and (46); we obtain
\[ 2 A \dot{A} h_{2,0}^0 e_{2,0} = A^2 \lambda_2^0 A h_{2,0}^0 e_{2,0} + A^2 Q(e_{1,0}, e_{1,0}) + O((A, A^*)^4), \]
so, using (39) and (40) together with the linear equation for \( A \), i.e., \( \dot{A} = \lambda A \), we obtain
\[ 2 \lambda h_{2,0}^0 = - \frac{4 \pi^2}{L_1^2} \left( 1 + \gamma \hat{V}_{2,0} \right) h_{2,0}^0 - \frac{8 \pi^2 L_2}{L_1} \gamma \hat{V}_{1,0}, \]
and finally, since \( \lambda_2^0 < 0 \), for \( \lambda \to 0^+ \) we formally get
\[ h_{2,0}^0 = - \frac{2 L_1 L_2 \gamma \hat{V}_{1,0}}{1 + \gamma \hat{V}_{2,0}}. \]
The reduced equation for \( A \) (47) then reads
\[ \dot{A} = \lambda A + 8 \gamma^2 \pi^2 L_2^2 \frac{\hat{V}_{1,0}}{1 + \gamma \hat{V}_{2,0}} \left( 2 \hat{V}_{2,0} - \hat{V}_{-1,0} \right) |A|^2 A. \]

From the assumptions of Theorem 6.1 and (37a) it follows that \( \hat{V}_{1,0} \) is negative, so if \( 2 \hat{V}_{2,0} - \hat{V}_{-1,0} > 0 \) the coefficient in front of the third order term is negative. This means that \( A(t) \) first grows exponentially, but then it saturates when the r.h.s. of (48) is equal to zero. This happens for
\[ |A| = \frac{\sqrt{\lambda}}{2 \sqrt{2 \gamma \pi L_2}} \frac{1 + \gamma \hat{V}_{2,0}}{\left| \frac{1}{\hat{V}_{1,0}} \left( 2 \hat{V}_{2,0} - \hat{V}_{-1,0} \right) \right|}. \]
Therefore, if the last factor is bounded, \( |A| \) is of order \( \sqrt{\lambda} \), so taking \( \lambda \) sufficiently small we ensure that \( A(t) \) remains small at the level of saturation, which justifies the validity of expansion (41).

When \( 2 \hat{V}_{2,0} - \hat{V}_{-1,0} < 0 \) the term of order \( A^3 \) does not bring any saturation. The growth thus goes on until \( A(t) \) leaves the perturbative regime, and at this point the approach breaks down.

This yields the hypothesis of Theorem 6.1. In order to conclude, we still need to justify that the manifold \( H \) can be represented by (43); we will prove the following lemma.
Lemma 6.2. Let $H = H[A, A^*](x)$ be as specified above in (41); then $\hat{H}_{0,0}[A, A^*] = 0$, $\hat{H}_{\pm 1,0}[A, A^*] = 0$ and the other Fourier coefficients of $H$ are of the form

$$\hat{H}_{k_1,k_2}[A, A^*] = \begin{cases} A^{k_1}h_{k_1,0}(\sigma) & \text{for } k_1 \geq 0, k_2 = 0, \\ (A^*)^{-k_1}h_{k_1,0}(\sigma) & \text{for } k_1 < 0, k_2 = 0, \\ 0 & \text{for } k_2 \neq 0, \end{cases}$$

for some unknown functions $h_{k_1,0} = h_{k_1,0}(\sigma)$, with $\sigma = AA^*$.

Proof. From the definition $\hat{H}_{0,0} = 0$, and $\hat{H}_{\pm 1,0} = 0$ since $\langle e_{1,0}, H \rangle = \langle e_{-1,0}, H \rangle = 0$. Next, (38) as well as the unstable manifold are invariant under translation $\tau_{x^0} : x \to x + x^0$ that act on functions as

$$(\tau_{x^0} \cdot f)(x) = f(x - x^0),$$

where $x = (x_1, x_2)$, $x^0 = (x_1^0, x_2^0)$. Therefore, for any $A$, there exists $\tilde{A}$ such that

$$\tau_{x^0} \cdot (Ae_{1,0} + A^* e_{-1,0} + H[A, A^*]) = \tilde{A}e_{1,0} + A^* e_{-1,0} + H[\tilde{A}, \tilde{A}^*],$$

meaning that

$$A e^{-i\pi \frac{x_1^0}{\tau_1}} e_{1,0} + A^* e^{i\pi \frac{x_1^0}{\tau_1}} e_{-1,0} + H[A, A^*](x - x_0) = \tilde{A}e_{1,0} + A^* e_{-1,0} + H[\tilde{A}, \tilde{A}^*](x).$$

Comparing the terms with $e_{1,0}$ we conclude that $\tilde{A} = Ae^{-i\pi \frac{x_1^0}{\tau_1}}$ and subsequently

$$H \left[ Ae^{-i\pi \frac{x_1^0}{\tau_1}}, A^* e^{i\pi \frac{x_1^0}{\tau_1}} \right](x) = H[A, A^*](x - x_0).$$

In terms of Fourier coefficients, the last equality reads

$$\hat{H}_{k_1,k_2} \left[ Ae^{-i\pi \frac{x_1^0}{\tau_1}}, A^* e^{i\pi \frac{x_1^0}{\tau_1}} \right] = e^{i\pi \left( \frac{k_1 x_1^0}{\tau_1} + \frac{k_2 x_2^0}{\tau_2} \right)} \hat{H}_{k_1,k_2}[A, A^*].$$

Let us now expand $\hat{H}_{k_1,k_2}$ in a Taylor series: $\hat{H}_{k_1,k_2}[z, z^*] = \sum_{l_1, l_2 \geq 0} c_{l_1,l_2} z^{l_1}(z^*)^{l_2}$; then (49) reads

$$\sum_{l_1, l_2 \geq 0} c_{l_1,l_2} A^{l_1}(A^*)^{l_2} e^{-i\pi \frac{x_1^0}{\tau_1} (l_1 - l_2)} = e^{i\pi \left( \frac{k_1 x_1^0}{\tau_1} + \frac{k_2 x_2^0}{\tau_2} \right)} \sum_{l_1, l_2 \geq 0} c_{l_1,l_2} A^{l_1} (A^*)^{l_2}.$$

The uniqueness of the expansion implies that $c_{l_1,l_2} = 0$ unless $l_1 - l_2 = k_1, k_2 = 0$. Thus

$$\hat{H}_{k_1,0}[A, A^*] = A^{k_1} \sum_{l_2 \geq 0} c_{k_1+l_2, l_2} A^{2l_2}.$$

This finishes the proof of Theorem 6.1. □
6.3. The square case for general potential—degenerate eigenvalues. In this section we study a particular case of domain—a periodic box; thus  $L_1 = L_2 = L$. For simplicity, we take  $L = \frac{1}{2}$. Again, the result is much more general and might be applied to a much wider class of functionals than the Hooke potential from section 5.3, provided one can select finitely many unstable modes. Here, due to the square symmetry, and assuming that the potential is isotropic, there will generically be one unstable mode in each direction denoted by

$$e_{1,0} = e^{2i\pi x_1} \quad \text{and} \quad e_{0,1} = e^{2i\pi x_2},$$

together with their conjugates, associated with the same eigenvalue,

$$\lambda = -4\pi^2 \left(1 + \gamma \hat{V}_{1,0}\right).$$

Our results in this case can be summarized as follows.

**Theorem 6.3.** Assume that $\lambda > 0$ and that $1 + \gamma \hat{V}_{k_1,k_2} > 0$ for any $k_1, k_2$ such that $|k_1| + |k_2| > 1$. Then, for

$$\frac{\hat{V}_{1,0}(2\hat{V}_{2,0} - \hat{V}_{-1,0})}{1 + \gamma \hat{V}_{2,0}} < -\frac{4 \hat{V}_{1,0}\hat{V}_{1,1}}{1 + \gamma \hat{V}_{1,1}}$$

the steady state exhibits a supercritical bifurcation. If the inequality is opposite, the steady state exhibits a subcritical bifurcation.

**Proof.** Following the same strategy as for the one-dimensional case we expand the perturbation $\eta$ on the unstable manifold:

$$\eta(t, x, y) = A(t)e_{1,0} + A^*(t)e_{-1,0} + B(t)e_{0,1} + B^*(t)e_{0,-1} + H[A, A^*, B, B^*](x, y);$$

therefore

$$\partial_t \eta(t, x, y) = \dot{A}e_{1,0} + \dot{A}^*e_{-1,0} + \dot{B}e_{0,1} + \dot{B}^*e_{0,-1} + \partial_t H[A, A^*, B, B^*](x, y).$$

Like in Lemma 6.2, we can deduce that $H$ has the following structure:

$$H = A^2e_{2,0}e_{2,0} + (A^2)^2e_{-2,0}e_{-2,0} + B^2e_{0,2}e_{0,2} + (B^2)^2e_{0,-2}e_{0,-2}$$

$$+ ABh_{1,1}e_{1,1} + A^*Bh_{-1,1}e_{-1,1} + AB^*h_{1,-1}e_{1,-1} + A^*B^*h_{-1,-1}e_{-1,-1}$$

$$+ O((A, A^*, B, B^*)^3).$$

We compute now the nonlinear term $N(\eta)$ at order $A^2, B^2$ (we use here the properties of $\hat{V}$: $\hat{V}_{k_1,k_2} = \hat{V}_{-k_1,-k_2} = \hat{V}_{-k_1,k_1} = \hat{V}_{k_2,k_1}$):

$$N(\eta) = -8\pi^2 \hat{V}_{1,0} \left[A^2e_{2,0} + B^2e_{0,2} + ABh_{1,1} + A^*Bh_{-1,1} + A^*Be_{1,-1} + c.c.\right]$$

$$+ O((A, A^*, B, B^*)^3).$$

The procedure is the same as before. The leading order for the dynamics of $A, B$ is the linear evolution:

$$\dot{A} = \lambda A + O((A, B)^3), \quad \dot{B} = \lambda B + O((A, B)^3).$$

We expand in powers of $\sigma_A = |A|^2, \sigma_B = |B|^2$ the $h_{k,l}$ coefficients that appear in (51) and keep only the leading order $h_{k,l}^0$, which are some constants to be computed. From
In order to check stability of the above steady states, we investigate the linearization of (2), solving the above equations, and letting $c < 0$, there are other steady states, with $A = A_{st} \neq 0$ and $B = B_{st} \neq 0$. The modulus of $A_{st}$ and $B_{st}$ is fixed, but their phase is undetermined:

$$|A_{st}| = |B_{st}| = \sqrt{\frac{\lambda}{-(c + d)}}.$$  

In order to check stability of the above steady states, we investigate the linearization of system (52), around $(A_{st}, B_{st})$. We take for simplicity $A_{st}$ and $B_{st}$ real in the...
following: by translation symmetry the result does not depend on the phases we choose. Furthermore, one checks easily that the linearized equations for the imaginary parts of $A$ and $B$ decouple from the real parts and are neutrally stable. We are left with the following linear equation for the real parts:

$$
\begin{bmatrix}
\hat{A} \\
\hat{B}
\end{bmatrix} = M(A_{st}, B_{st}) \begin{bmatrix}
A \\
B
\end{bmatrix}, \quad M(A_{st}, B_{st}) = \lambda \begin{pmatrix}
1 - \frac{3c + d}{c + d} & \frac{2d}{c + d} \\
\frac{2d}{c + d} & 1 - \frac{3c + d}{c + d}
\end{pmatrix}.
$$

The eigenvalues of $M(A_{st}, B_{st})$ are equal to $\xi_1 = -2$, $\xi_2 = 2\sqrt{\frac{c}{c+d}}$, and so the steady state is stable if $c < d$. This, together with the condition $c + d < 0$, implies that the system (52) possesses a stable steady state provided $c < -|d|$ as assumed in (50). Otherwise, the steady state is unstable.

### 6.4. Numerical tests for the Hookean potential.

We now compute the values of parameters $c$ and $d$ (53) for various values of parameters $\alpha$ and $\beta$ corresponding to the slightly unstable case (close to the instability threshold). For simplicity we consider the case of unit periodic box, i.e., $L_1 = L_2 = L = \frac{1}{2}$, so that (33) gives

$$
\hat{V}_{k_1, k_2} = \frac{2\pi R^4}{\tilde{\pi}^2 k_1^2 k_2^2} \left( \frac{\pi \alpha}{2} \left[ J_1(z_{k_1, k_2})H_0(z_{k_1, k_2}) - J_0(z_{k_1, k_2})H_1(z_{k_1, k_2}) \right] - J_2(z_{k_1, k_2}) \right),
$$

where $z_{k_1, k_2} = 2\pi R \sqrt{j^2 + k^2}$, $\alpha = \frac{\pi R}{\tilde{R}}$. Since we are in the periodic box, we know from Proposition 5.5 that the instability appears for larger values of parameter $\beta$ than in the whole space case, i.e., for $\beta > \beta_c = \frac{3\pi}{4\sqrt{2}}$.

The assumptions of Theorem 6.3 are met if

$$
1 + \gamma \hat{V}_{1,0} = 1 + \frac{\beta}{(2\pi R)^2} \left( \frac{\pi \alpha}{2} \left[ J_1(2\pi R)H_0(2\pi R) - J_0(2\pi R)H_1(2\pi R) \right] - J_2(2\pi R) \right) < 0
$$

and

$$
1 + \gamma \hat{V}_{1,1} = 1 + \frac{\beta}{(2\pi R)^2} \left( \frac{\pi \alpha}{2} \left[ J_1(2\sqrt{2}\pi R)H_0(2\sqrt{2}\pi R) - J_0(2\sqrt{2}\pi R)H_1(2\sqrt{2}\pi R) \right] - J_2(2\sqrt{2}\pi R) \right) > 0.
$$

Note that according to the definition of function $F^{\alpha,\beta}$ (28) the above conditions are equivalent to

$$
F^{\alpha,\beta}(2\pi R) < 0, \quad F^{\alpha,\beta}(2\sqrt{2}\pi R) > 0,
$$

and from the proof of Proposition 5.4 we know that the rest of the eigenvalues in the assumption of Theorem 6.3 will have a good sign as well.

We will now present computations of coefficients $c$ and $d$ defined in (53) that are used in Theorem 6.3 to determine the condition for the type of bifurcation (50). To this purpose we choose parameter $\alpha$ in the unstable regime, where $\alpha = \frac{1}{2}$, and for several values of $R \leq L = \frac{1}{2}$ we first find the critical value of parameter $\beta$, for which the bifurcation occurs. Having this parameter we compute $c$ and $d$ using the expressions (53) in which we take $\gamma = \frac{2\pi}{\sqrt{2}\pi R}$, then criterion (50) to identify the type of bifurcation. Our results are summarized in Table 1.
The numerical results for the rectangular domain $[-1/2,1/2] \times [-1/2,1/2]$ for three different values of interaction radius $R$. The corresponding critical values of the parameter $\beta$ from the second column are computed using the condition $\lambda = 0$. The character of the bifurcation in the third column is identified using criterion (50).

<table>
<thead>
<tr>
<th>$R$</th>
<th>$\beta_c$</th>
<th>Type of transition</th>
</tr>
</thead>
<tbody>
<tr>
<td>$1/2$</td>
<td>83.0</td>
<td>continuous</td>
</tr>
<tr>
<td>$1/4$</td>
<td>31.1</td>
<td>discontinuous</td>
</tr>
<tr>
<td>$1/8$</td>
<td>25.5</td>
<td>discontinuous</td>
</tr>
</tbody>
</table>

These computations are in line with the analysis in [14], according to which for short-range potentials (when $R/L$ is small), the transition tends to become discontinuous (first order), which corresponds to the subcritical dynamical scenario. Note that the present bifurcation analysis provides a precise criterion for the boundary between the first order/subcritical and second order/supercritical cases.

Numerical results addressing the comparison between the microscopic and macroscopic approach can be found in recent work [2].

Data statement. No new data was collected in the course of this research.

Acknowledgments. The first author thanks the Department of Mathematics at Imperial College for hospitality. The third author wishes to thank José Antonio Carrillo for suggesting the literature and for stimulating discussions on the subject.

REFERENCES


