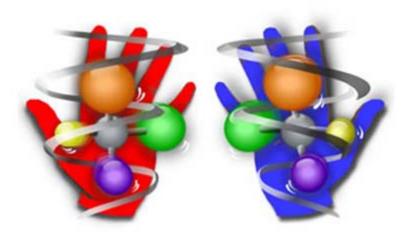




Synopsis: Noise Gives Biology a Hand

October 8, 2015

The preferred handedness of biomolecules may be the result of noisy chemical fluctuations, according to a new theory.



NPG Asia Materials

Many molecules essential to life come with a given handedness, or chirality: amino acids are always left handed, while sugars are always right handed. Why this "homochirality" exists, however, is a mystery, since it takes a similar amount of energy to form left- and right-handed versions of a molecule. Farshid Jafarpour and colleagues at the University of Illinois at Urbana–Champaign have now suggested that the asymmetry might be the consequence of noisy fluctuations accompanying the chemical reactions that produce these biomolecules.

According to current theories, homochirality emerged through several steps. First, some environmental asymmetry—arising, for instance, from the Earth's rotation or exposure to circularly polarized light—created a small imbalance between the two species. The imbalance was then amplified as each chiral species catalyzed its own production (through self-replication). However, these two steps are not sufficient to explain biology's

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perfect selectivity, so researchers have hypothesized that a chiral species becomes dominant by also inhibiting the formation of the other. But recent experiments cast doubts on this hypothesis by showing that two chiral variants of RNA can coexist.

Jafarpour and colleagues provide a simpler explanation that does not rely on cross-inhibition. They analyze a set of self-replication reactions that produce the two chiral species and calculate how their ratio evolves in the presence of concentration fluctuations—a form of noise arising because chemical reactions are probabilistic processes. Their calculations show that, even if the two species were initially equally probable, fluctuations move the system inexorably toward a homochiral state. Once reached, this state is stable because self-replicating reactions can only produce molecules with the same chirality.

This research is published in **Physical Review Letters**.

-Matteo Rini

Subject Areas

Biological Physics

Physical Chemistry

Noise-Induced Mechanism for Biological Homochirality of Early Life Self-Replicators

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Noise-Induced Mechanism for Biological Homochirality of Early Life Self-Replicators

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The observed single-handedness of biological amino acids and sugars has long been attributed to autocatalysis. However, the stability of homochiral states in deterministic autocatalytic systems relies on cross inhibition of the two chiral states, an unlikely scenario for early life self-replicators. Here, we present a theory for a stochastic individual-level model of autocatalysis due to early life self-replicators. Without chiral inhibition, the racemic state is the global attractor of the deterministic dynamics, but intrinsic multiplicative noise stabilizes the homochiral states, in both well-mixed and spatially extended systems. We conclude that autocatalysis is a viable mechanism for homochirality, without imposing additional nonlinearities such as chiral inhibition.

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One of the very few universal features of biology is homochirality: every naturally occurring amino acid is left-handed (L chiral) while every sugar is right-handed (D chiral) [1,2]. Although such unexpected broken symmetries are well known in physics—for example, in the weak interaction—complete biological homochirality still defies explanation. In 1953, Charles Frank suggested that homochirality could be a consequence of chemical autocatalysis [3], frequently presumed to be the mechanism associated with the emergence of early life self-replicators. Frank introduced a model in which the D and L enantiomers of a chiral molecule are autocatalytically produced from an achiral molecule A in reactions $A + D \rightarrow 2D$ and $A + L \rightarrow 2L$, and the enantiomers are consumed in a chiral inhibition reaction, $D + L \rightarrow 2A$ [4]. The state of this system can be described by the chiral order parameter ω defined as $\omega \equiv (d-l)/(d+l)$, where d and l are the concentrations of D and L. The order parameter ω is zero at the racemic state, and ± 1 at the homochiral states. Frank's model has three deterministic fixed points of the dynamics; the racemic state is an unstable fixed point, and the two homochiral states are stable fixed points. Starting from almost everywhere in the D-L plane, the system converges to one of the homochiral fixed points [Fig. 1(a)].

In the context of biological homochirality, extensions of Frank's idea have essentially taken two directions. On the one hand, the discovery of a synthetic chemical system of amino alcohols that amplifies an initial excess of one of the chiral states [5] has motivated several autocatalysis-based models (see Ref. [6] and the references therein). On the other hand, ribozyme-driven catalyst experiments [7] have inspired theories based on polymerization and chiral inhibition that minimize [8–10] or do not include at all [11,12] autocatalysis. In contrast, a recent experimental

realization of ribonucleic acid replication using a novel ribozyme shows such efficient autocatalytic behavior that chiral inhibition does not arise [13]. Further extensions accounting for both intrinsic noise [6,14] and diffusion [15–18] build further upon Frank's work.

Regardless of the specific model details, all of these models share the three-fixed-points paradigm of Frank's model, namely, that the time evolution of the chiral order parameter ω is given by a deterministic equation of the form [6]

$$\frac{d\omega}{dt} = f(t)\omega(1 - \omega^2),\tag{1}$$

where the function f(t) is model dependent. However, the homochiral states arise from a nonlinearity which is not a property of simple autocatalysis but, for instance in the original Frank's model, is due to chiral inhibition [see Fig. 1(b)]. The sole exception to the three-fixed-points model in a variation of Frank's model is the work of Lente [19], where purely stochastic chiral symmetry breaking occurs, although chiral symmetry breaking is only partial, with $\omega \neq 0$ but $|\omega| < 1$.

The purpose of this Letter is to show that efficient early life self-replicators can exhibit universal homochirality, through a stochastic treatment of Frank's model *without* requiring nonlinearities such as chiral inhibition. In our stochastic treatment, the homochiral states arise not as fixed points of deterministic dynamics, but instead are states where the effects of chemical number fluctuations (i.e., the multiplicative noise [20]) are minimized. The mathematical mechanism proposed here [21–24] is intrinsically different from that of the class of models summarized by Eq. (1). In the following, we propose a model which we analytically solve for the spatially uniform case and the case of two

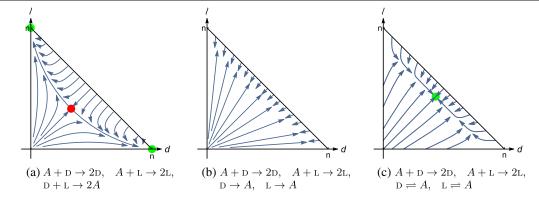


FIG. 1 (color online). (a) Phase portrait of Frank's model: the racemic state is an unstable fixed point (the red dot), while the homochiral states are stable fixed points (the green dots). (b) If chiral inhibition is replaced by a linear decay reaction, the ratio of D and L molecules stays constant. (c) Adding even the slightest amount of nonautocatalytic production of D and L molecules makes the racemic state (the green dot) the global attractor of the dynamics.

well-mixed patches coupled by diffusion. We then show, using numerical simulations, that the results persist in a one-dimensional spatially extended system. We conclude that autocatalysis alone can, in principle, account for universal homochirality in biological systems.

Stochastic model for well-mixed system.—Motivated in part by the experimental demonstration of autocatalysis without chiral inhibition [13], we propose the reaction scheme below, which is equivalent to a modification of Lente's reaction scheme [19] through the additional process representing the recycling of enantiomers:

$$A + D \xrightarrow{k_a} 2D$$
, $A + L \xrightarrow{k_a} 2L$,
 $A \xrightarrow{k_n} D$, $A \xrightarrow{k_n} L$. (2)

Compared to Frank's model, the chiral inhibition is replaced by linear decay reactions which model both recycling and nonautocatalytic production. The rate constants are denoted by k, with the subscript serving to identify the particular reaction. The only deterministic fixed point of this model is the racemic state [Fig. 1(c)]. This model can be interpreted as a model of the evolution of early life where primitive chiral self-replicators can be produced randomly through nonautocatalytic processes at very low rates; the self-replication is modeled by autocatalysis, while the decay reaction is a model for the death process [25].

We now approximate reaction scheme (2) by means of a stochastic differential equation for the time evolution of the chiral order parameter, ω , which shows that in the regime where autocatalysis is the dominant reaction, the functional form of the multiplicative intrinsic noise from autocatalytic reactions stabilizes the homochiral states. We consider a well-mixed system of volume V and total number of molecules N. As shown in the Supplemental Material (SM) [26], for $N \gg 1$, we obtain the following equation for ω , defined in the Itō sense [20]:

$$\frac{d\omega}{dt} = -\frac{2k_n k_d V}{Nk_a} \omega + \sqrt{\frac{2k_d}{N} (1 - \omega^2)} \eta(t), \qquad (3)$$

where $\eta(t)$ is normalized Gaussian white noise [20].

The time-dependent distribution of Eq. (3) can be computed exactly [24,27]. The stationary distribution [20],

$$P_s(\omega) = \mathcal{N}(1 - \omega^2)^{\alpha - 1}, \quad \text{with} \quad \alpha = \frac{Vk_n}{k_a}, \quad (4)$$

depends on a single parameter, α , where the normalization constant $\mathcal N$ is given by

$$\mathcal{N} = \left(\int_{-1}^{+1} (1 - \omega^2)^{\alpha - 1} d\omega \right)^{-1} = \frac{\Gamma(\alpha + \frac{1}{2})}{\sqrt{\pi} \Gamma(\alpha)}. \tag{5}$$

Equation (4) is compared in Fig. 2 against Gillespie simulations [28] of scheme (2). For $\alpha = \alpha_c = 1$, ω is

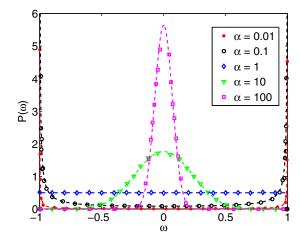


FIG. 2 (color online). Comparison between the stationary distribution, Eq. (4) (dashed lines), and Gillespie simulations of the reactions (2) (markers), for different values of α . Simulation parameters: $N = 10^3$, $k_a = k_n = k_d = 1$.

uniformly distributed. For $\alpha \gg \alpha_c$, where the nonautocatalytic production is the dominant production reaction, $P_s(\omega)$ is peaked around the racemic state, $\omega=0$. For $\alpha\ll\alpha_c$, where autocatalysis is dominant, $P_s(\omega)$ is sharply peaked around the homochiral states, $\omega=\pm 1$. The simulations were performed for N=1000, where the analytic theory is expected to be accurate; for smaller values of N, the theory is qualitatively correct, but very small quantitative deviations are observable compared to the simulations. For example, for $N\sim 100$, $\alpha_c\sim 1.005$.

The deterministic part of Eq. (3) has one fixed point at the racemic state, consistent with the phase portrait in Fig. 1(c). The multiplicative noise in Eq. (3) vanishes at homochiral states and admits its maximum at the racemic state. For $\alpha \ll \alpha_c$, where autocatalysis is dominant, the amplitude of the noise term in Eq. (3) is much larger than the amplitude of the corresponding deterministic term. In this regime, the system ends up at homochiral states where the noise vanishes.

To understand this result physically, note that the source of the multiplicative noise is the intrinsic stochasticity of the autocatalytic reactions. While, on average, the two autocatalytic reactions do not change the variable ω , each time one of the reactions takes place, the value of ω changes by a very small discrete amount. As a result, over time the value of ω drifts away from its initial value. Since the amplitude of the noise term is maximum at racemic state and zero at homochiral states, this drift stops at one of the homochiral states. The absence of the noise from autocatalysis at homochiral states can be understood by recognizing that at homochiral states, the molecules with only one of two chiral states D and L are present; hence, only the autocatalytic reaction associated with that chiral state has a nonzero rate. This reaction produces molecules of the same chirality, keeping the system at the same homochiral state without affecting the value of ω , and, therefore, the variable ω does not experience a drift away from the homochiral states due to the autocatalytic reactions.

Since the stationary distribution of ω in Eq. (4) is only dependent on α , the decay reaction rate, k_d has no effect on the steady state distribution of the system. The only role of this reaction is to prevent the A molecules from being completely consumed, thus providing a well-defined nonequilibrium steady state independent of the initial conditions. The parameter α is proportional to the ratio of the nonautocatalytic production rate, k_n , to the self-replication rate, k_a . In the evolution of early life, when self-replication was a primitive function, k_a would be small and the value of α would therefore be large; however, as self-replication became more efficient, the value of k_a would have increased and thus α would have decreased. Therefore, in our model, we expect that life started in a racemic state, and it transitioned to complete homochirality through the mechanism explained above, after self-replication became efficient (i.e., when $\alpha \ll \alpha_c$).

It is important to note that all of the previous mechanisms suggested for homochirality rely on assumptions that cannot be easily confirmed to hold during the emergence of life. However, even if all of such mechanisms fail during the origin of life, our mechanism guarantees the emergence of homochirality since it only relies on self-replication and death, two processes that are inseparable from any living system.

Stochastic model with spatial extension.—We now turn to the study of reaction scheme (2) generalized to the spatially extended case [29]. We discretize space into a collection of M patches of volume V, indexed by i. The geometry of the space is defined by $\langle i \rangle$ —the set of patches that are nearest neighbor to patch i (e.g., for a linear chain, $\langle i \rangle = \{i-1,i+1\}$). We indicate the molecules of species A in patch i by A_i and similarly for the other species. Each patch is well mixed and reactions (2) occur within, while molecules can diffuse between neighboring patches with diffusion rate δ . In summary, the following set of reactions defines the spatial model:

$$A_{i} \stackrel{k_{n}}{\rightleftharpoons} D_{i}, \qquad A_{i} \stackrel{k_{n}}{\rightleftharpoons} L_{i}, \qquad i = 1, ..., M,$$

$$A_{i} + D_{i} \stackrel{k_{a}}{\rightleftharpoons} 2D_{i}, \qquad A_{i} + L_{i} \stackrel{k_{a}}{\rightleftharpoons} 2L_{i},$$

$$D_{i} \stackrel{\delta}{\rightleftharpoons} D_{j}, \qquad L_{i} \stackrel{\delta}{\rightleftharpoons} L_{j}, \qquad j \in \langle i \rangle.$$

$$(6)$$

We now derive the following set of coupled stochastic differential equation for the time evolution of the chiral order parameter ω_i , of each patch i (see the SM [26])

$$\frac{d\omega_i}{dt} = -\frac{2k_n k_d V}{Nk_a} \omega_i + \delta \sum_{j \in \langle i \rangle} (\omega_j - \omega_i)
+ \sqrt{\frac{2k_d}{N} (1 - \omega_i^2)} \eta_i(t) + \sqrt{\frac{\delta}{N}} \xi_i(\vec{\omega}, t), \quad (7)$$

where now N represents the average number of molecules per patch, η_i 's are independent normalized Gaussian white noises, ξ_i 's are zero mean Gaussian noise with correlator

$$\langle \xi_i(t)\xi_j(t')\rangle = \left(2\sum_{k\in\langle i\rangle} (1-\omega_i\omega_k)\delta_{i,j} + (\omega_i^2 + \omega_j^2 - 2)\chi_{\langle i\rangle}(j)\right)\delta(t-t'), \quad (8)$$

and $\chi_{(i)}(j)$ is equal to one if $j \in \langle i \rangle$ and zero otherwise.

In order to see how the coupling of well-mixed patches affects their approach to homochirality, it is instructive to consider the simplest case of two adjacent patches (M=2). In the two-patch model, various scenarios can happen: the system may not exhibit homochirality ($\omega_1 \sim \omega_2 \sim 0$); each patch can separately reach homochirality ($\omega_1 = \pm 1$ and $\omega_2 = \pm 1$); the system exhibits global homochirality ($\omega_1 = \omega_2 = \pm 1$). We first analyze the condition for each patch reaching homochirality using perturbation theory, in

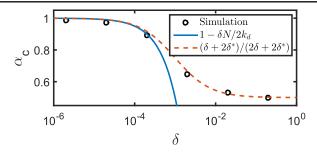


FIG. 3 (color online). Parameter $\alpha_c^{\rm patch}$ in the two-patch system as a function of the diffusion rate δ . Gillespie simulations (markers) are compared against Eq. (11) (the solid blue line) and Eq. (13) (the dashed red line). Simulation parameters as in Fig. 2.

the case of slow diffusion. The stationary probability density function of the chiral order parameter of a single patch, $Q_s(\omega)$, is defined by

$$Q_s(\omega) = \int_{-1}^{+1} Q_s(\omega, \omega_2) d\omega_2 = \int_{-1}^{+1} Q_s(\omega_1, \omega) d\omega_1, \quad (9)$$

where $Q_s(\omega_1, \omega_2)$ is the joint probability distribution of ω_1 and ω_2 at steady state from Eq. (7). If $\delta \sim k_d/N$ or smaller, then (see the SM [26]) the stationary distribution reads

$$Q_s(\omega) = Z(1 - \omega^2)^{\alpha + (\delta N/2k_d) - 1},$$
 (10)

where \mathcal{Z} is a normalization constant. This result shows that the critical α in a single patch, up to the first order correction in δ , is given by

$$\alpha_c^{\text{patch}} \approx 1 - \delta \frac{N}{2k_d}, \quad \text{for } \delta \approx 0.$$
 (11)

We can now turn to the case of high diffusion. Recall that the patches are defined as the maximum volume around a point in space in which the system can be considered well mixed. This can be interpreted as the maximum volume in which diffusion dominates over the other terms acting on the variable of interest (in this case ω). From Eq. (7), this condition is fulfilled for $\delta \sim 2k_d\alpha/N$. In the vicinity of the transition, α is on the order of one; therefore, the condition becomes $\delta \sim k_d/N$. For $\delta \gg k_d/N$, the whole system can be considered well mixed, and we can find the critical value of α for each patch, starting from $\alpha_c = 1$, from the well-mixed results, and using as volume the volume of the whole system, i.e., MV. This indicates that, in a single patch,

$$\alpha_c^{\text{patch}} \approx \frac{1}{M}, \quad \text{for } \delta \gg 0.$$
 (12)

A simple formula that interpolates between these extreme limits, asymptotic to 1/M (with M=2) for large δ and to Eq. (11) for small δ , is

$$\alpha_c^{\text{patch}} = \frac{\delta + 2\delta^*}{2\delta + 2\delta^*}, \qquad \delta^* = \frac{k_d}{N}.$$
 (13)

Figure 3 shows agreement between α_c^{patch} , measured from Gillespie simulations of the two-patch system, and Eq. (13). At the parameter regime below the α_c curve in Fig. 3, individual patches are homochiral. Also, we find that the correlation between the homochiral states of the two patches increases with diffusion rate δ , and the patches become completely correlated when $\delta \sim k_d/N$ or more. In this regime the system reaches global homochirality.

This latter result suggests that in the spatially extended model, when autocatalysis is the dominant reaction (i.e., α is small enough) and when the diffusion rate is in the order of k_d/N or larger, all patches converge to the same homochiral state. Figure 4 shows the dynamics of a Gillespie simulation of a one-dimensional chain of 100 patches, initializes at the racemic state, in the pure autocatalytic limit ($k_n \rightarrow 0$). Very quickly, small islands of different homochirality (blue and red) are formed. Islands of opposite chirality compete against each other, until the system reaches global homochirality. Note that for

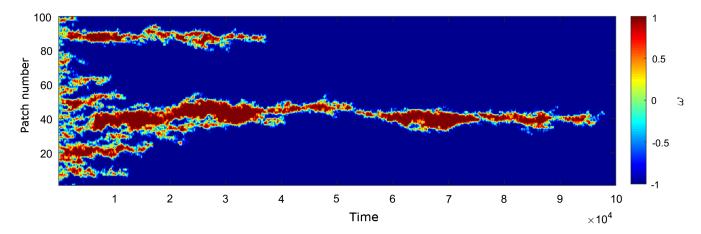


FIG. 4 (color online). Gillespie simulation of scheme (6) for a one-dimensional system of M=100 patches, starting from the racemic state and ending with all of the patches in the same homochiral state, $\omega=-1$. Simulation parameters: N=1000, $k_a=k_d=1$, $\delta=10^{-3}$, and $k_n=0$.

 $\delta \sim k_d/N$ we can treat the diffusion process deterministically by ignoring the last term in Eq. (7). In this regime, Eq. (7) is the same as the equation describing the one-dimensional voter model, implying that the transition to homochirality is in the universality class of compact directed percolation [30].

In conclusion, a racemic population of self-replicating chiral molecules far from equilibrium, even in the absence of other nonlinearities that have previously been invoked, such as chiral inhibition, transitions to complete homochirality when the efficiency of self-replication exceeds a certain threshold. This transition occurs due to the drift of the chiral order parameter under the influence of the intrinsic stochasticity of the autocatalytic reactions. The functional form of the multiplicative intrinsic noise from autocatalysis directs this drift toward one of the homochiral states. Unlike some other mechanisms in the literature, this process does not require an initial enantiomeric excess. In our model, the homochiral states are not deterministic dynamical fixed points but are instead stabilized by intrinsic noise. Moreover, in the spatial extension of our model, we have shown that diffusively coupled autocatalytic systems synchronize their final homochiral states, allowing a system solely driven by autocatalysis to reach global homochirality. We conclude that autocatalysis alone is a viable mechanism for homochirality, without the necessity of imposing chiral inhibition or other nonlinearities.

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- [26] See Supplemental Material at http://link.aps.org/supplemental/10.1103/PhysRevLett.115.158101, which includes Refs. [20, 26, 27, 30, 31], for derivation of the Eq. (2) and (7) from the individual level models and the derivation of Eq. (10) using perturbation theory.
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Supplementary Material: A noise-induced mechanism for biological homochirality of early life self-replicators

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(Dated: August 16, 2015)

WELL-MIXED SYSTEM

We start from reaction scheme (2) of the main paper

$$A + D \xrightarrow{k_a} 2D, A + L \xrightarrow{k_a} 2L,$$

$$A \xrightarrow[k_d]{k_a} D, A \xrightarrow[k_d]{k_a} L.$$
(S1)

Each reaction changes the system from a state $\vec{x} \equiv (x_1, x_2, x_3) \equiv (a, d, l)$, specified by the concentration of molecules A, D, and L, to a state of the form $\vec{x} + V^{-1}\vec{s}_m$ (for some $m \in \{1, \ldots, 4\}$), where \vec{s}_m is the m'th row of the stoichiometry matrix

$$\mathbf{S} = \begin{pmatrix} -1 & 1 & 0 \\ -1 & 0 & 1 \\ 1 & -1 & 0 \\ 1 & 0 & -1 \end{pmatrix}. \tag{S2}$$

The probability per unit time of such transition is given by transition rates $T(\vec{x} + V^{-1}\vec{s}_m|\vec{x})$ obtained from law of mass action for reactions (S1):

$$T(\vec{x} + \frac{1}{V}\vec{s}_1|\vec{x}) = V(k_n + k_a d)a, \quad T(\vec{x} + \frac{1}{V}\vec{s}_3|\vec{x}) = Vk_d d,$$

$$T(\vec{x} + \frac{1}{V}\vec{s}_2|\vec{x}) = V(k_n + k_a l)a, \quad T(\vec{x} + \frac{1}{V}\vec{s}_4|\vec{x}) = Vk_d l.$$
(S3)

The set of rates (S3) is used to write the master equation for the time evolution of the probability density function, $P(\vec{x},t)$, of the system being in the state \vec{x} at time t [1].

We begin by defining the functions F_m 's as

$$F_m(\vec{x},t) = T(\vec{x}|\vec{x} - \frac{1}{V}\vec{s}_m)P(\vec{x} - \frac{1}{V}\vec{s}_m,t),$$
 (S4)

so that the master equation can be written as:

$$\frac{\partial P(\vec{x},t)}{\partial t} = -\sum_{m=1}^{4} \left(F_m(\vec{x} + \frac{1}{V}\vec{s}_m, t) - F_m(\vec{x}, t) \right).$$
 (S5)

This equation defines the stochastic model and can be numerically simulated using the Gillespie algorithm [2].

In order to initiate an analytical treatment, we begin by expanding the right-hand side of the master equation (we follow [3]). We obtain a Gaussian noise approximation by truncating the expansion at the second-order, thus neglecting terms corresponding to higher moments. We arrive at the non-linear Fokker-Planck equation:

$$\frac{\partial P}{\partial t} \approx -\sum_{j=1}^{3} \frac{\partial (H_j P)}{\partial x_j} + \frac{1}{2} \sum_{j,k=1}^{3} \frac{\partial^2 (B_{jk} P)}{\partial x_j \partial x_k}, \quad (S6)$$

where the drift vector \vec{H} with component H_i reads:

$$\vec{H} = \frac{1}{V} \sum_{m} T(\vec{x} + \frac{1}{V} \vec{s}_{m}) |\vec{x}| \vec{s}_{m}$$

$$= \begin{pmatrix} k_{d}(d+l) - a(2k_{n} + k_{a}(d+l)) \\ -k_{d}d + a(k_{n} + k_{a}d) \\ -k_{d}l + a(k_{n} + k_{a}l) \end{pmatrix}.$$
(S7)

The symmetric diffusion matrix has the form

$$\mathbf{B} = \frac{1}{V^2} \sum_{m} T(\vec{x} + \frac{1}{V} \vec{s}_m) |\vec{x}| (\vec{s}_m \otimes \vec{s}_m) = \frac{1}{V} \begin{pmatrix} k_d(d+l) + a(2k_n + k_a(d+l)) & -k_dd - a(k_n + k_ad) & -k_dl - a(k_n + k_al) \\ -k_dd - a(k_n + k_ad) & k_dd + a(k_n + k_ad) & 0 \\ -k_dl - a(k_n + k_al) & 0 & k_dl + a(k_n + k_al) \end{pmatrix},$$
(S8)

where the symbol \otimes indicates the Kronecker product. We now decompose the diffusion matrix to $\mathbf{B} = \mathbf{G}\mathbf{G}^T$. Multiple choices for \mathbf{G} exist [3], and it is easy to check that

the following 3×2 matrix satisfies the decomposition:

$$\mathbf{G} = \frac{1}{\sqrt{V}} \begin{pmatrix} \sqrt{a(k_a d + k_n) + k_d d} & \sqrt{a(k_a l + k_n) + k_d l} \\ -\sqrt{a(k_a d + k_n) + k_d d} & 0 \\ 0 & -\sqrt{a(k_a l + k_n) + k_d l} \end{pmatrix}.$$

Equation (S6) is equivalent to the following stochastic differential equation (defined in the Itō sense) [4]

$$\frac{\mathrm{d}\vec{x}}{\mathrm{d}t} = \vec{H}(\vec{x}) + \mathbf{G}(\vec{x})\vec{\eta}(t) \tag{S10}$$

where η_k 's (k = 1, 2) are Gaussian white noises with zero mean and correlation

$$\langle \eta_i(t)\eta_k(t')\rangle = \delta_{ik}\delta(t-t').$$
 (S11)

Note that since, the Fokker-Planck equation (S6) only depends on **B** and not the particular choice of its decomposition **G**, the probability density function of \vec{x} and its time evolution do not depend on **G** either [3].

The number of degrees of freedom in Eq. (S10) can be reduced by noting two facts: (i) the reaction scheme (S1) conserves the total number of molecules, meaning that the total concentration n=a+d+l is conserved; (ii) simulations show that the concentration r=d+l settles to a Gaussian distribution around its fixed point value r^* , allowing us to substitute $r(t) \to r^*$. We therefore change variables in Eq. (S10) (using Itō's formula) [4], from

$$\begin{pmatrix} a \\ d \\ l \end{pmatrix} \to \begin{pmatrix} n \\ r \\ \omega \end{pmatrix} = \begin{pmatrix} a+d+l \\ d+l \\ (d-l)/(d+l) \end{pmatrix}, \quad (S12)$$

so that the only dynamics occurs in the chiral order parameter ω . In the new variables, we find that $\dot{n}=0$ and, by taking the positive solution of $\dot{r}=0$, that is

$$r^* = \frac{\sqrt{(k_a n - k_d - 2k_n)^2 + 8k_a k_n n} + k_a n - k_d - 2k_n}{2k_a},$$
(S13)

we substitute $r \to r^*$ in the equation for ω , and use the rule for summing Gaussian variables (i.e. $a\eta_1 + b\eta_2 = \sqrt{a^2 + b^2}\eta$; where a and b are generic functions [4]) to express the stochastic part of the equation using a single noise variable. Expressing the result in terms of the total number of molecules N = Vn, for $N \gg 1$, we arrive at the following stochastic differential equation for chirality order parameter ω (equation (2) of the main text):

$$\frac{\mathrm{d}\omega}{\mathrm{d}t} = -\frac{2k_n k_d V}{Nk_a}\omega + \sqrt{\frac{2k_d}{N}(1-\omega^2)}\eta(t), \quad (S14)$$

where $\eta(t)$ is Gaussian white noise with zero mean and unit variance. The corresponding Fokker-Planck equation of Eq. (S14) is an exactly solvable partial differential equation with time dependent solution given in [5]. The steady state probability distribution of ω is given by

$$P_s(\omega) = \mathcal{N} (1 - \omega^2)^{\alpha - 1}, \text{ with } \alpha = \frac{V k_n}{k_a}, \text{ (S15)}$$

with the normalization constant

$$\mathcal{N} = \left(\int_{-1}^{+1} \left(1 - \omega^2 \right)^{\alpha - 1} d\omega \right)^{-1} = \frac{\Gamma\left(\alpha + \frac{1}{2} \right)}{\sqrt{\pi} \Gamma(\alpha)}. \quad (S16)$$

TWO-PATCH MODEL

Starting from reaction scheme (6) of the main paper for the spatial extension of our model

$$A_{i} \xrightarrow{k_{a}} D_{i}, \quad A_{i} \xrightarrow{k_{a}} L_{i}, \quad i = 1, \dots, M$$

$$A_{i} + D_{i} \xrightarrow{k_{a}} 2D_{i}, \quad A_{i} + L_{i} \xrightarrow{k_{a}} 2L_{i}$$

$$D_{i} \xrightarrow{\delta} D_{j}, \quad L_{i} \rightleftharpoons L_{j}, \quad j \in \langle i \rangle,$$
(S17)

for M=2, we can follow the procedure explained in the previous section to obtain a Fokker-Planck equation for time evolution of the probability density of system being at a state with concentrations $a_1,\ d_1,\ l_1,\ a_2,\ d_2,\ {\rm and}\ l_2.$ Again we can reduce the number of variables using the following facts (i) the total concentration $n_t=n_1+n_2=a_1+d_1+l_1+a_2+d_2+l_2$ is conserved; (ii) simulation shows that in long time, the variables $r_1=d_1+l_1,\ r_2=d_2+l_2,$ and $\Delta=n_1-n_2$ settle to Gaussian distributions around their fixed point values $r_1=r_2=r^*$ and $\Delta=0$. We do the following change of variables

$$\begin{pmatrix} a_1 \\ d_1 \\ l_1 \\ a_2 \\ d_2 \\ l_2 \end{pmatrix} \rightarrow \begin{pmatrix} n_t \\ \Delta \\ r_1 \\ r_2 \\ \omega_1 \\ \omega_2 \end{pmatrix} = \begin{pmatrix} a_1 + d_1 + l_1 + a_2 + d_2 + l_2 \\ a_1 + d_1 + l_1 - a_2 - d_2 - l_2 \\ d_1 + l_1 \\ d_2 + l_2 \\ (d_1 - l_1)/(d_1 + l_1) \\ (d_2 - l_2)/(d_2 + l_2) \end{pmatrix}$$
(S18)

using Itō's formula. Now the dynamics only occurs only in $\vec{\omega} = (\omega_1, \omega_2)$. For large average number of molecules per patch $N \gg 1$, the resulting Fokker-Planck equation for time evolution of the joint probability density function of ω_1 and ω_2 , $Q(\vec{\omega}, t)$, reads

$$\frac{\partial Q}{\partial t} = -\sum_{i=1}^{2} \frac{\partial ((\mathbf{L}\vec{\omega})_{i} Q)}{\partial \omega_{i}} + \frac{1}{2} \sum_{i,j=1}^{2} \frac{\partial^{2} (U_{ij} Q)}{\partial \omega_{i} \partial \omega_{j}}.$$
 (S19)

Note that the above sums are now over the patches, and not over species as in Eq. (S6). The Jacobian matrix L

$$\mathbf{L} = -\frac{2k_d k_n V}{Nk_a} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + \delta \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix}, \quad (S20)$$

and the diffusion matrix ${\bf U}$ reads

$$\mathbf{U} = \frac{2k_d}{N} \begin{pmatrix} 1 - \omega_1^2 & 0\\ 0 & 1 - \omega_2^2 \end{pmatrix} + \frac{\delta}{N} \begin{pmatrix} 2(1 - \omega_1 \omega_2) & \omega_1^2 + \omega_2^2 - 2\\ \omega_1^2 + \omega_2^2 - 2 & 2(1 - \omega_1 \omega_2). \end{pmatrix}.$$
 (S21)

Note that the stochastic differential equation corresponding to Eq. (S19) is the M=2 case of equation (7) of the main paper.

Perturbation theory for the small diffusion case

Now, we can use perturbation theory for small δ , to find the stationary probability density of ω for each patch defined as

$$Q_s(\omega) = \int_{-1}^{+1} Q_s(\omega, \omega_2) d\omega_2 = \int_{-1}^{+1} Q_s(\omega_1, \omega) d\omega_1,$$
(S22)

where $Q_s(\omega_1, \omega_2)$ is the stationary solution of Eq. (S19). For $\delta \sim k_d/N$ or smaller, we can treat the diffusion deterministically by ignoring the last term in Eq. (S21). To solve for $Q_s(\omega)$, we begin by rewriting Eq. (S19) as a continuity equation,

$$\partial_t Q + \nabla \cdot \vec{J} = 0, \tag{S23}$$

which defines the probability current \vec{J} as [4]

$$\vec{J} = \mathbf{L}\vec{\omega} \ Q - \frac{1}{2}\nabla \cdot (\mathbf{U}Q) \,. \tag{S24}$$

By the conservation of probability, at stationary conditions, the total probability flux \vec{J}_s through each vertical section of ω_1 - ω_2 plane must be zero. That is

$$\int_{-1}^{+1} J_{s,1} d\omega_2 = \int_{-1}^{+1} \left((\mathbf{L}\vec{\omega})_1 Q_s - \frac{1}{2} \partial_{\omega_1} (U_{11} Q_s) \right) d\omega_2$$

$$= Q_s(\omega_1) \omega_1 \left(\frac{2k_d}{N} (1 - \alpha) - \delta \right) - \frac{k_d}{N} (1 - \omega_1^2) \frac{dQ_s}{d\omega_1}$$

$$+ \delta \int_{-1}^{+1} \omega_2 Q_s(\omega_1, \omega_2) d\omega_2 = 0.$$
(S25)

The last integral can be evaluated using Bayes' theorem

$$\delta \int_{-1}^{+1} \omega_2 Q_s(\omega_1, \omega_2) d\omega_2 = \delta \int_{-1}^{+1} \omega_2 Q_s(\omega_2 | \omega_1) Q_s(\omega_1) d\omega_2$$
$$= \delta Q_s(\omega_1) \langle \omega_2 \rangle_{\omega_1} = \mathcal{O}(\delta^2), \tag{S26}$$

which is of order δ^2 for small δ , since, $\langle \omega_2 \rangle_{\omega_1}$ (the expected value of ω_2 given ω_1) vanishes at zero δ , and therefore, of order δ for small δ . In this regime, Eq. (S25) provide us with a differential equation for $Q_s(\omega)$ with the solution (equation (10) of the main paper)

$$Q_s(\omega) = \mathcal{Z}(1 - \omega^2)^{\alpha + \frac{\delta N}{2k_d} - 1}, \tag{S27}$$

where the normalization constant \mathcal{Z} is given by

$$\mathcal{Z} = \frac{\Gamma\left(\alpha + \frac{\delta N}{2k_d} + \frac{1}{2}\right)}{\sqrt{\pi} \Gamma\left(\alpha + \frac{\delta N}{2k_d}\right)}.$$
 (S28)

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