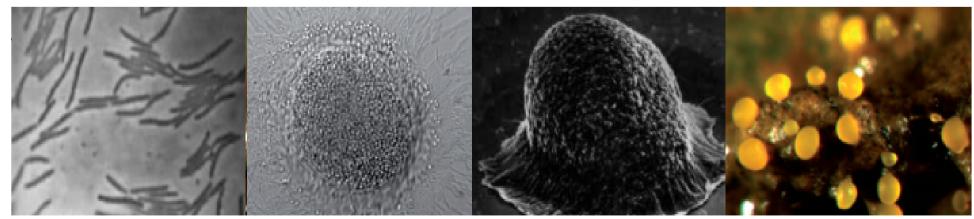




Active soft-matter: modeling living matter and mimicking it.

Fernando Peruani

BIOMAT – La Falda – 08.2014













outline

P1

- 1. Definition of active soft-matter
- 2. Examples in biology and non-living systems
- 3. Minimal models of active particles
 - a. non-interacting
 - b. interacting
- 4. Symmetries and boundary conditions
- 5. Gas-liquid-like transitions

outline

- 1. Definition of active soft-matter
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P2

What is active soft-matter?

It is a novel area of research at the interface between biology, mathematics, and soft-matter physics that is being developed to understand the physics behind non-equilibrium systems such as bacterial colonies, tissue formation, cancer growth, and more generally, embryogenesis.

Active soft-matter (ASM) has recently witnessed the emergence of a promising new direction: the design and construction of biomimetic, active materials, which are essentially ensembles of artificial active particles.

Specifically, ASM deals with ensembles of active particles and their macroscopic properties.

What is an active particle?

What is the difference between a passive and an active particle?

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What is an active particle?

It is a particle able to convert energy into work in order to self-propel in a dissipative medium.

(for instance, it could be a particle equipped with a propelling engine.)

What is the difference between a passive and an active particle?

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What is the difference between a passive and an active particle?

There is energy consumption, energy dissipation, while the FDT does not necessary apply, and in general, momentum (neither linear nor angular) is not conserved.

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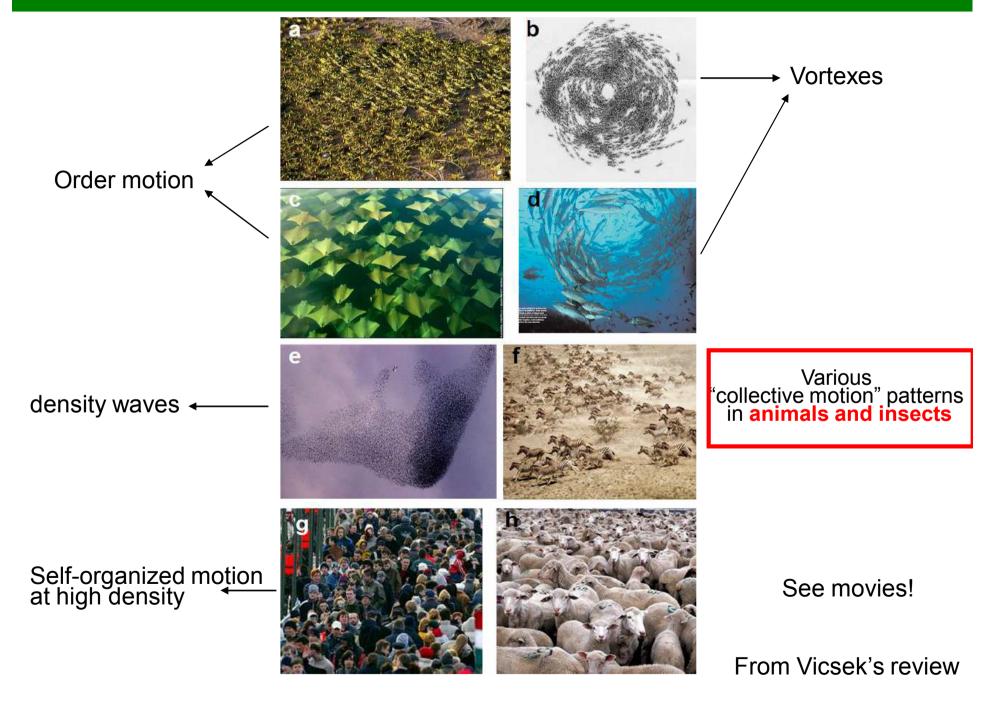
There is energy consumption, energy dissipation, while the FDT does not necessary apply, and in general, momentum (neither linear nor angular) is not conserved.

Are there examples of such kind of particles?

Plenty!

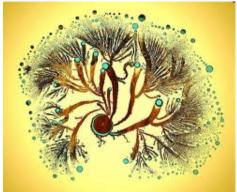
Active matter systems: examples

Pattern formation and active matter



Pattern formation and active matter

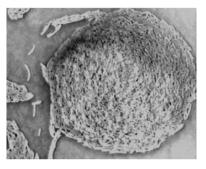


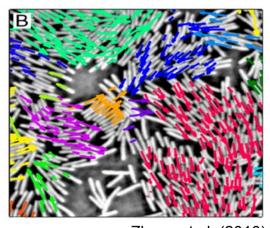


Various CM patterns in bacteria

Vortex in bacteria!

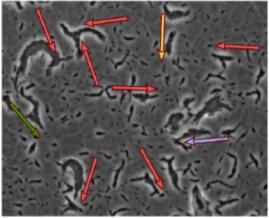
Paenibacillus Ben-Jacob's group





Zhang et al. (2010)

B. subtilis



Peruani et al. (2011)

Myxobacteria

Order motion and clustering In bacteria

Myxobacteria

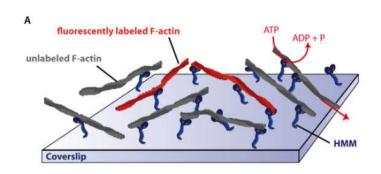


Strassmann and Queller (2011)

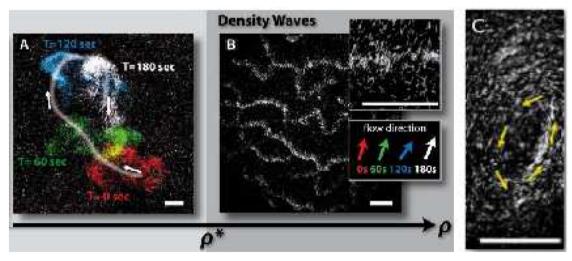
Super complex coordinated motion in bacteria: They pile up and form complex 3D structures!

[See movie from Reichenbach (1965) and others]

Pattern formation and active matter

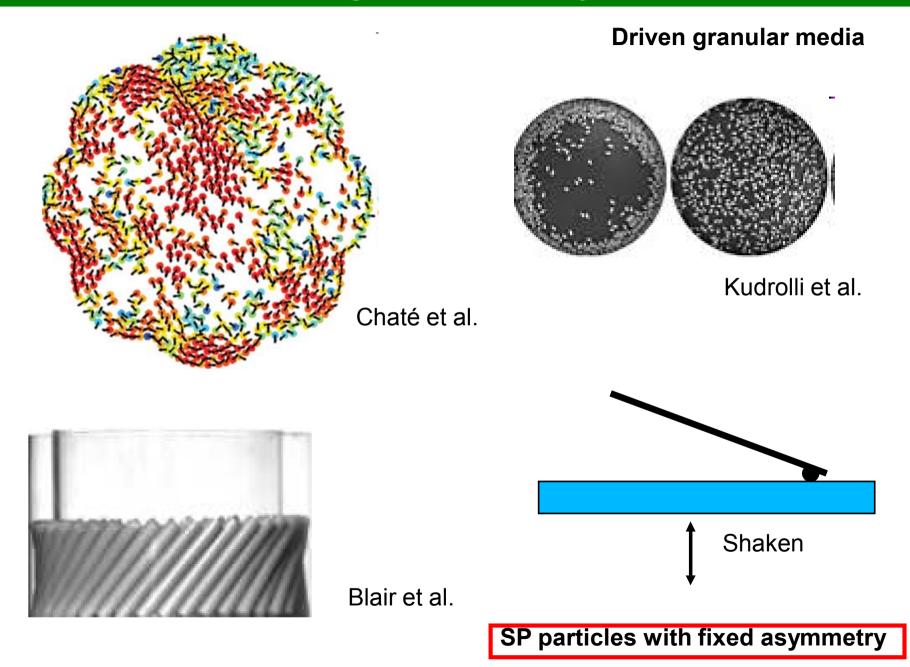


Various CM patterns at intracellular scale



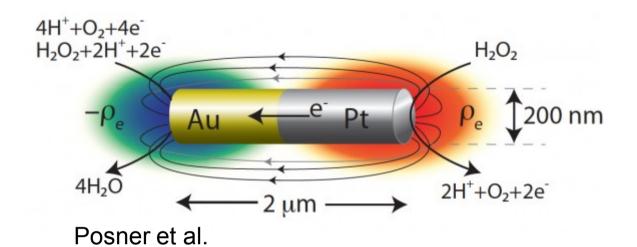
Schaller et al. (2010)

Non-living active matter systems



Non-living active matter systems

Chemically driven particles



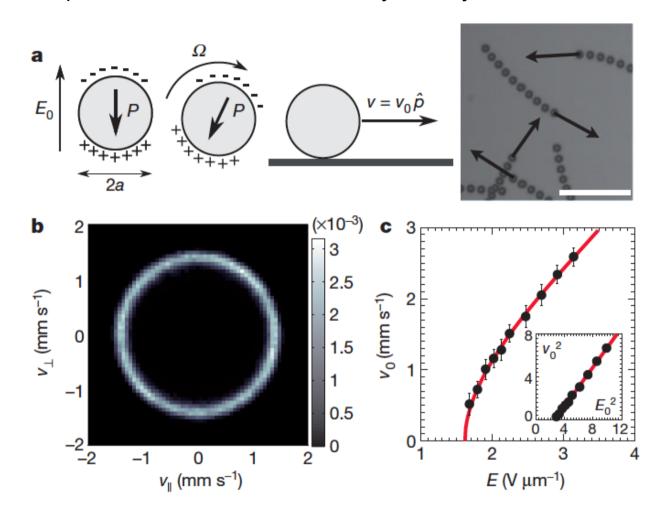


Freemantle et al.

SP particles with fixed asymmetry

Non-living active matter systems

Experimental example that exhibits the claimed symmetry: Quincke rollers!



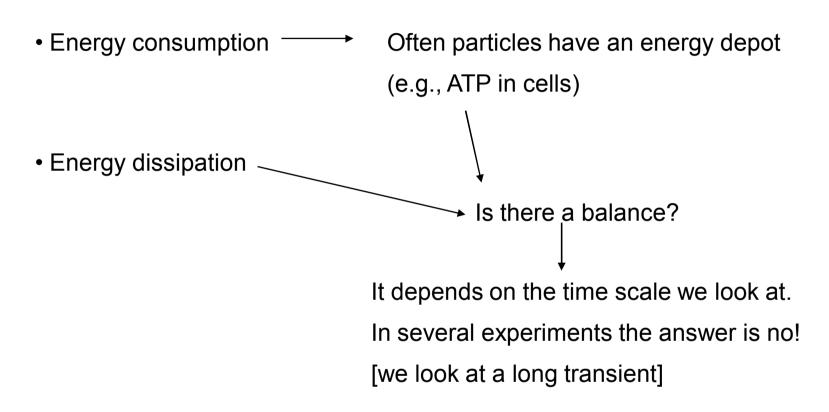
SP particles that undergo a spontaneous symmetry breaking

Bricard, Caussin, Desreumaux, Dauchot, Bartolo, Nature (2013)

Active matter systems: theoretical challenges & minimal models

Theoretical challenges

Active systems are intrinsically non-equilibrium systems



- Often active systems (if we forget about the medium where they move) do not conserve momentum!
- Anomalous fluctuations are found everywhere: density, speed, etc, etc

Theoretical challenges

active systems - classification

- Non-interacting active particles
- Interacting active particles
 - long-range interactions
 - short-range interactions
 - homogeneous populations of particles
 - inhomogeneous populations of particles (e.g., with leaders)

We will focus (mainly) on identical active particles with short-range interactions

Theoretical challenges

Active systems - classification

• Interaction "forces":

- Attraction
- Repulsion
- Alignment (velocity-velocity interactions!)

A minimal model for non-interacting active particles:

Equations of motion:

$$\dot{\mathbf{x}}_i = v_i(t)\mathbf{V}(\theta_i)$$
 $\dot{\theta}_i = \eta \xi_i(t)$

where
$$V(\theta) \equiv (\cos(\theta), \sin(\theta))$$

Fluctuations in the direction of motion and in the <u>propelling engine!</u>

Random direction of motion + constant speed:



If you drive a car at constant speed while randomly turning the direction wheel, you perform a well-known type of motion called persistent Brownian motion.

Random direction of motion + fluctuations in the speed:





If now instead of keeping the speed constant, you randomly press the gas and break pedals while still randomly turning the direction wheel, the resulting movement is no longer well described by a plain persistent Brownian motion.

Fluctuations in the direction of motion + fluctuations in the speed

=

Anomalous (Brownian) motion

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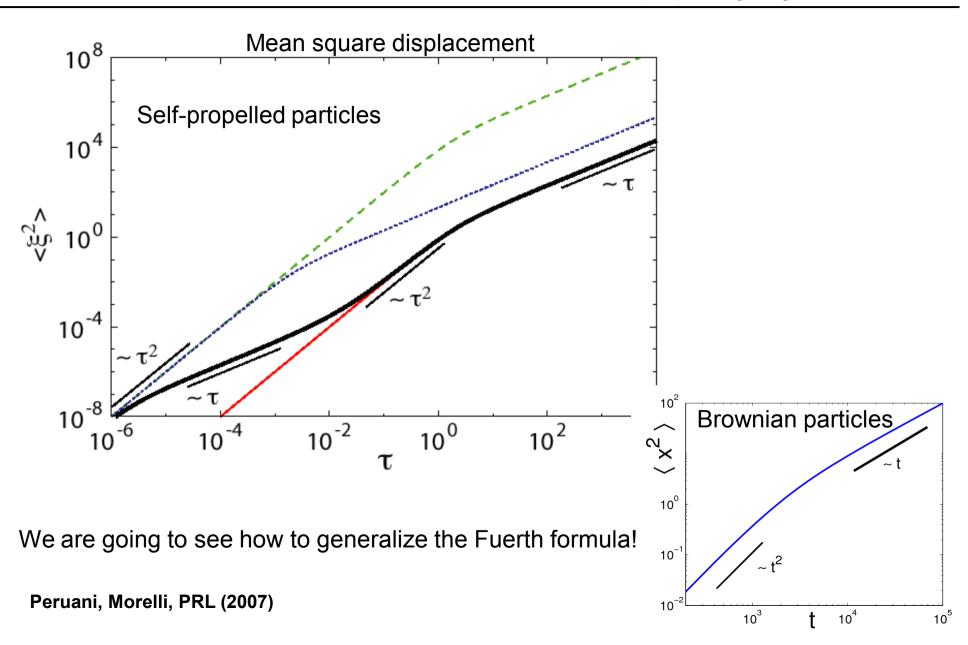
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Fluctuations in the direction of motion + fluctuations in the speed

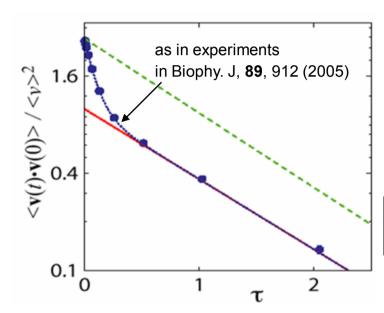
=

Anomalous (Brownian) motion

Fluctuations in the direction of motion and in the propelling engine!



Active fluctuations and the anomalous transient can be observed in cell motility experiments



$$\langle \mathbf{x}^{2}(t) \rangle = 2 \frac{\langle v \rangle^{2}}{\kappa^{2}} (\kappa t - 1 + e^{-\kappa t})$$

$$+ 2 \frac{\langle v^{2} \rangle - \langle v \rangle^{2}}{(\kappa + \beta)^{2}} ((\kappa + \beta)t - 1 + e^{-(\kappa + \beta)t}).$$

$$D = D_{Thermal} + D_{Active} + D_{Active Fluct}$$

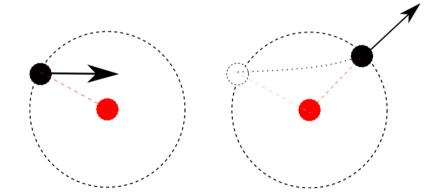
A simple model of active particles in heterogeneous media:

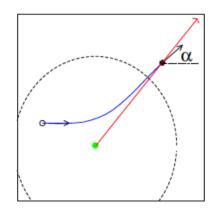
Equations of motion:

$$egin{array}{lll} \dot{\mathbf{x}}_i &= v_0 \mathbf{V}(\theta_i) \ \dot{ heta}_i &= h(\mathbf{x}_i) + \eta \xi_i(t) \end{array}$$
 where $\mathbf{V}(heta) \equiv (\cos(heta), \sin(heta))$

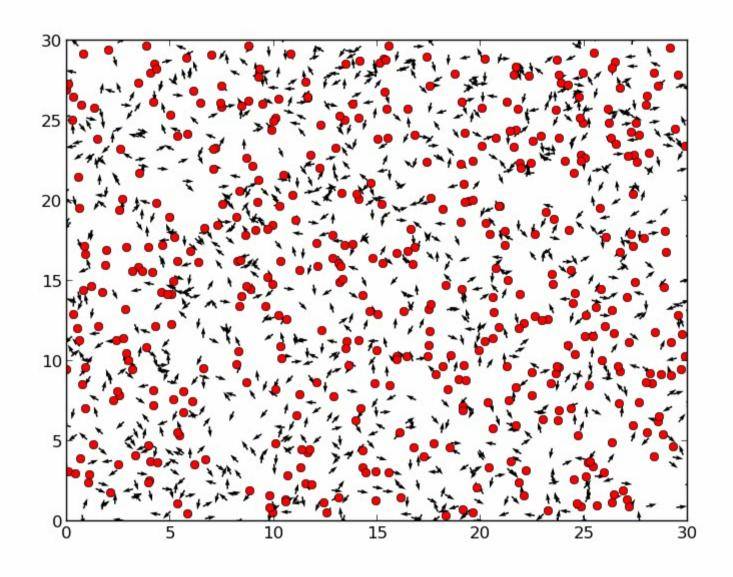
Interaction with "obstacles":

$$h(\mathbf{x}_i) = \begin{cases} \frac{\gamma}{n(\mathbf{x}_i)} \sum_{\Omega_i} \sin(\alpha_{k,i} - \theta_i) & \text{if } n(\mathbf{x}_i) > 0\\ 0 & \text{if } n(\mathbf{x}_i) = 0 \end{cases}$$

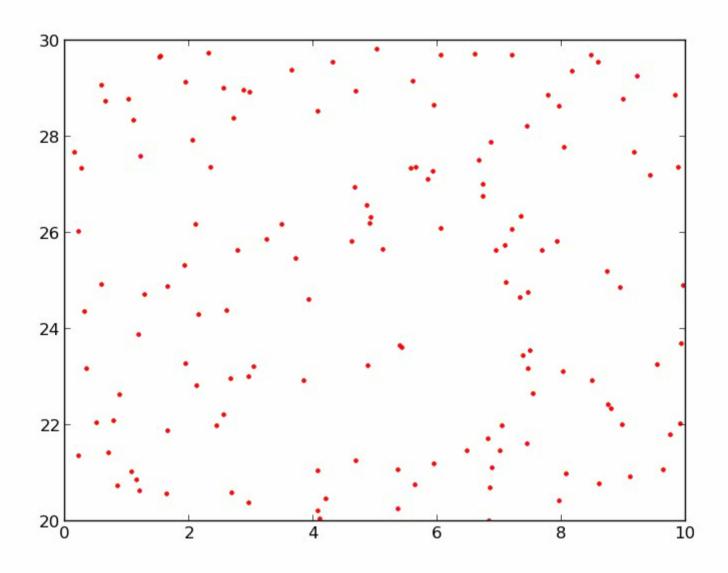




O. Chepizhko, F. Peruani, PRL (2013)

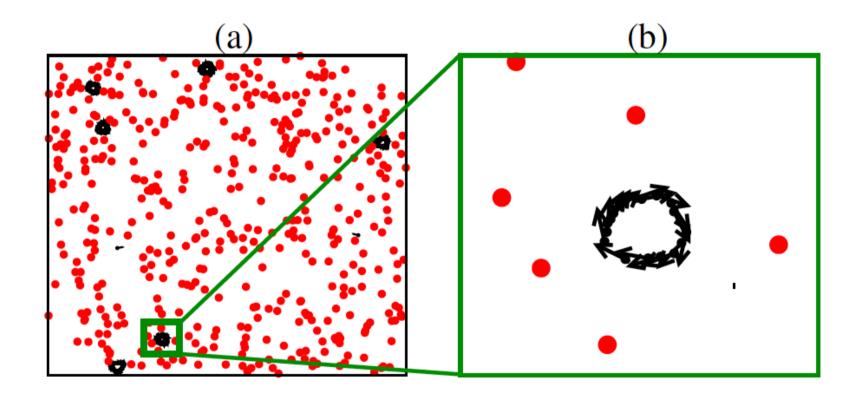


Starting from a random initial condition, traps emerge. The time that particles spend in a particular trap depends on the particular configuration of obstacles.



Looking at a small part of a large system at real time

• How is possible to trapped particles moving at constant speed?

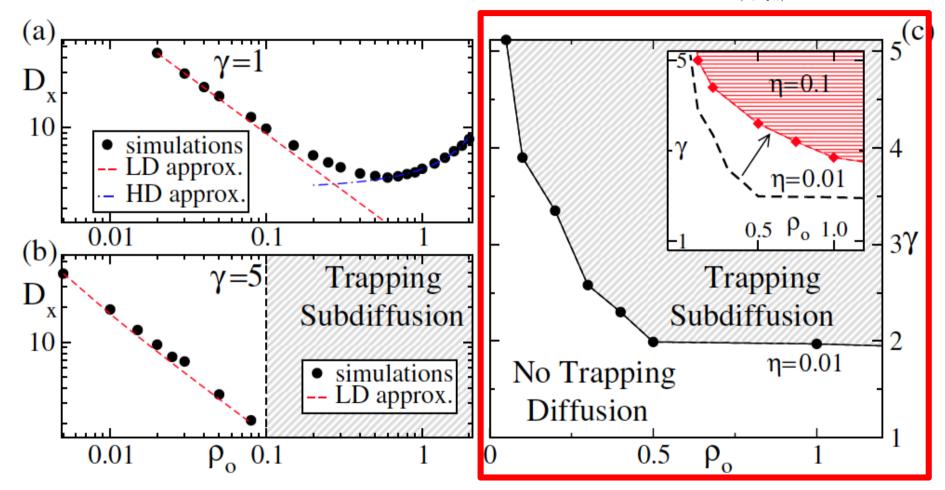


These traps are closed stable orbits that are found by the active particles in the landscape of obstacles.

Remember that noise is present. These orbits are not absorbing, and particle can escape.

• The transport properties – dependency with \mathbf{y} and $\mathbf{\rho}_{\underline{\mathbf{o}}}$ $\dot{\theta_i} = \frac{\gamma}{n(\mathbf{x}_i)} \sum_{|\mathbf{x}_i - \mathbf{y}_k| < R} \sin(\alpha_{k,i} - \theta_i) + \eta \xi_i(t)$

$$\dot{\mathbf{x}}_i = v_0 \mathbf{V}(\theta_i)$$
 $\dot{\theta}_i = \frac{\gamma}{n(\mathbf{x}_i)} \sum_{|\mathbf{x}_i - \mathbf{y}_k| < R} \sin(\alpha_{k,i} - \theta_i) + \eta \xi_i(t)$



Diffusion and subdiffusion are observed!

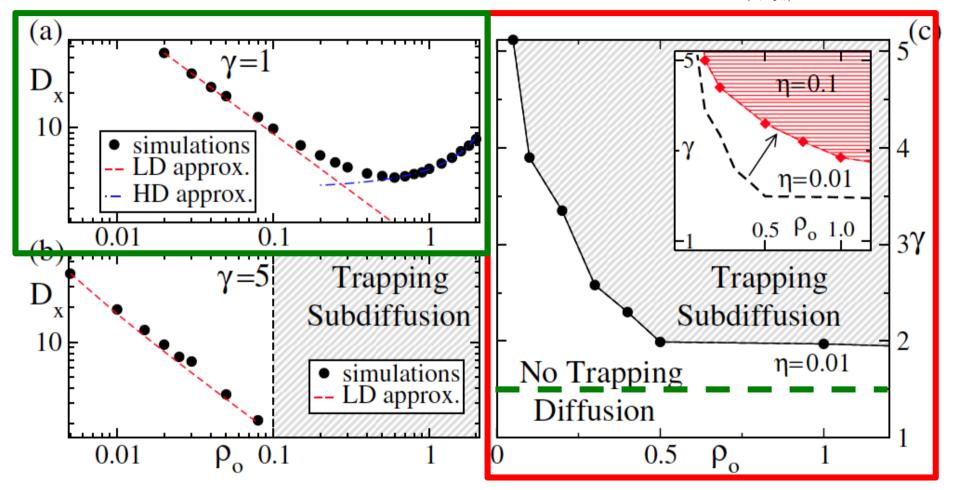
Subdiffusion is related to the emergence of trapping!

Chepizhko, Peruani, PRL (2013)

• The transport properties – dependency with γ and $\rho_{\underline{o}}$ $\dot{\theta}_i = \frac{\gamma}{n(\mathbf{x}_i)}$

$$\dot{\mathbf{x}}_i = v_0 \mathbf{V}(\theta_i)$$

$$\dot{\theta}_i = \frac{\gamma}{n(\mathbf{x}_i)} \sum_{|\mathbf{x}_i - \mathbf{y}_k| < R} \sin(\alpha_{k,i} - \theta_i) + \eta \xi_i(\theta_i)$$



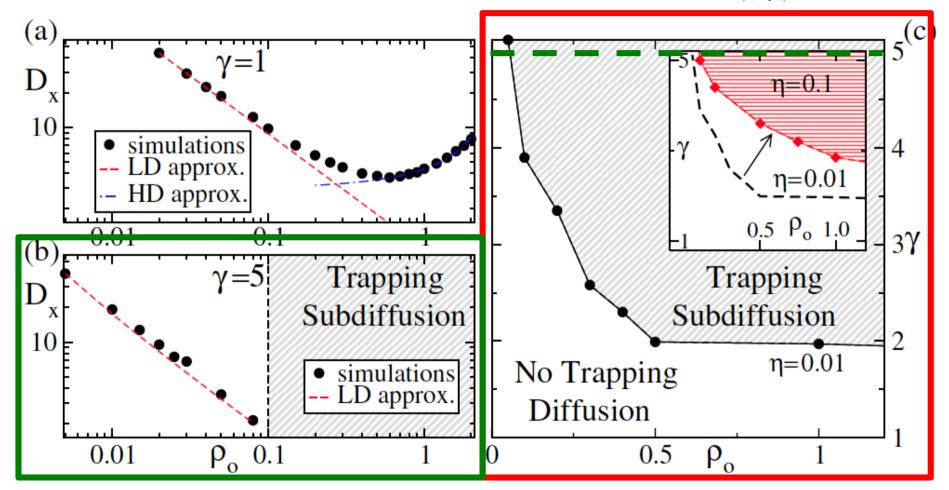
Diffusion and subdiffusion are observed!

Subdiffusion is related to the emergence of trapping!

(These findings can be understood theoretically!)

• The transport properties – dependency with \mathbf{y} and $\mathbf{\rho}_{\underline{\mathbf{o}}}$ $\dot{\theta_i} = \frac{\gamma}{n(\mathbf{x}_i)} \sum_{|\mathbf{x}_i - \mathbf{y}_k| < R} \sin(\alpha_{k,i} - \theta_i) + \eta \xi_i(t)$

$$\dot{\mathbf{x}}_i = v_0 \mathbf{V}(\theta_i)$$
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Diffusion and subdiffusion are observed!

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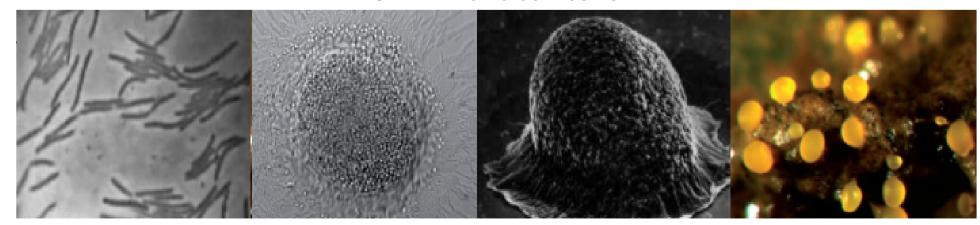


Active soft-matter: modeling living matter and mimicking it.

-- second part --

Fernando Peruani

BIOMAT – La Falda – 08.2014



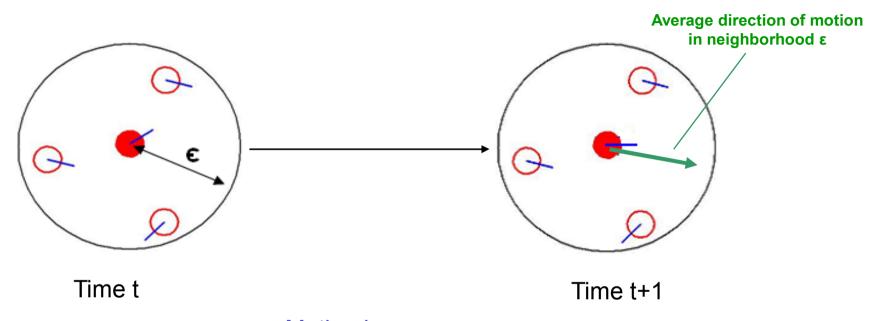
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P2

The Vicsek model – moving spins

[T. Vicsek et al., Phys. Rev. Lett. 75, 1226 (1995)]



Motion in space

$$\mathbf{x}_i(t+1) = \mathbf{x}_i(t) + \mathbf{v}_i(t)\Delta t$$

Change of the direction of motion

$$\theta(t+1) = \langle \theta(t) \rangle_r + \Delta \theta$$

Average direction of motion at time t

Angular noise !!!

A more general, better defined active particle models

$$\dot{\mathbf{x}}_i = v_0 e^{\mathrm{i}\,\theta_i}$$

$$\dot{\mathbf{x}}_i = v_0 e^{i \theta_i}$$

$$\dot{\theta}_i = -\gamma \frac{\partial U}{\partial \theta_i} (\mathbf{x}_i, \theta_i) + \tilde{\eta}_i(t)$$

Ferromagnetic alignment
$$\rightarrow U(\theta, \theta') = -\cos(\theta - \theta')$$

Polar order parameters:

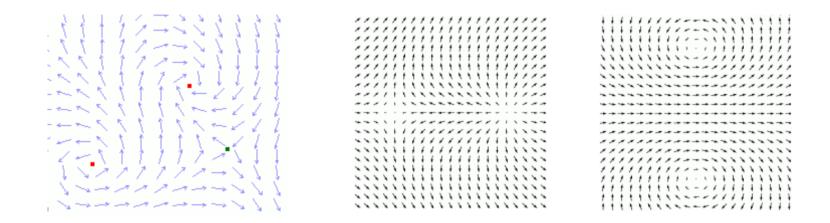
$$\phi = \langle \left| \frac{1}{N} \sum_{k=1}^{N} \exp(\mathrm{i} \, \theta_k^t) \right| \rangle \longrightarrow \begin{array}{l} \text{[global] average velocity} \\ \text{(or average magnetization if we treat the velocity vector as spins).} \end{array}$$

[Peruani, Deutsch, and Bär, EPJ ST 157, 111 (2008)]

Imperfect flow of information leads to defects, and defects set the limit to the size/scale of the patterns we can observe

Some classical results from statistical mechanics:

• The Kosterlitz-Thouless transition: $E_{ij} = -J(\mathbf{s}_i \mathbf{s}_j) = -J\cos(\varphi_i - \varphi_j)$



• The Mermin-Wagner theorem:

In equilibrium systems with SU2 symmetry, long-range order out of short-range interactions cannot emerge in 1D or 2D!

What is short- and long-range order? And how to measure it?

Imperfect flow of information leads to defects, and defects set the limit to the size/scale of the patterns we can observe

The Mermin-Wagner theorem:

In equilibrium systems with SU2 symmetry, long-range order out of short-range interactions cannot emerge in 1D or 2D!

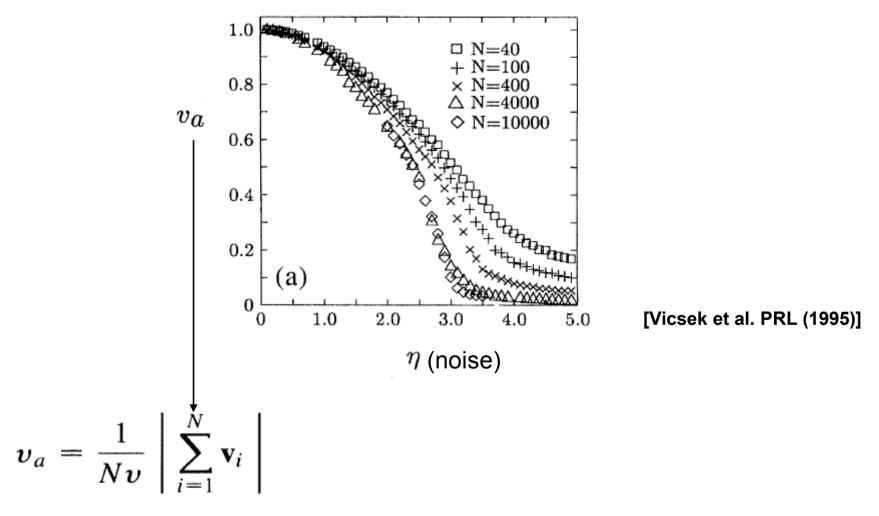
- Long-range order is not possible in equilibrium systems (spin waves and vortices)
- MW theorem also works in several non-equilibrium system
- Random motion (of diffusive type) of the spins is not enough to get long-range order in the thermodynamical limit (though it is for finite systems)

Properties of the Vicsek model – when the spins move!



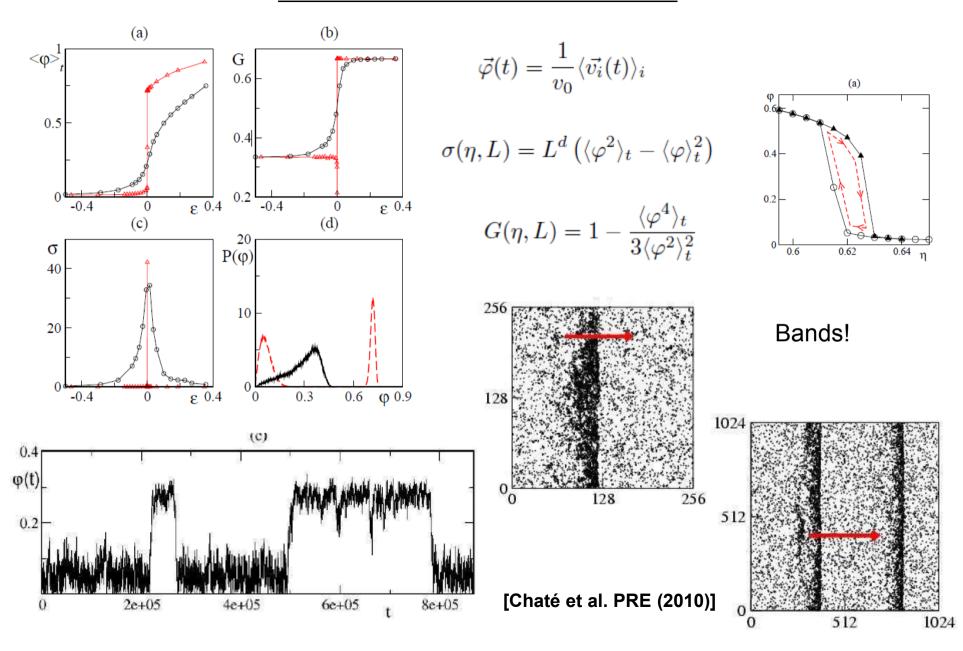
Decreasing noise values

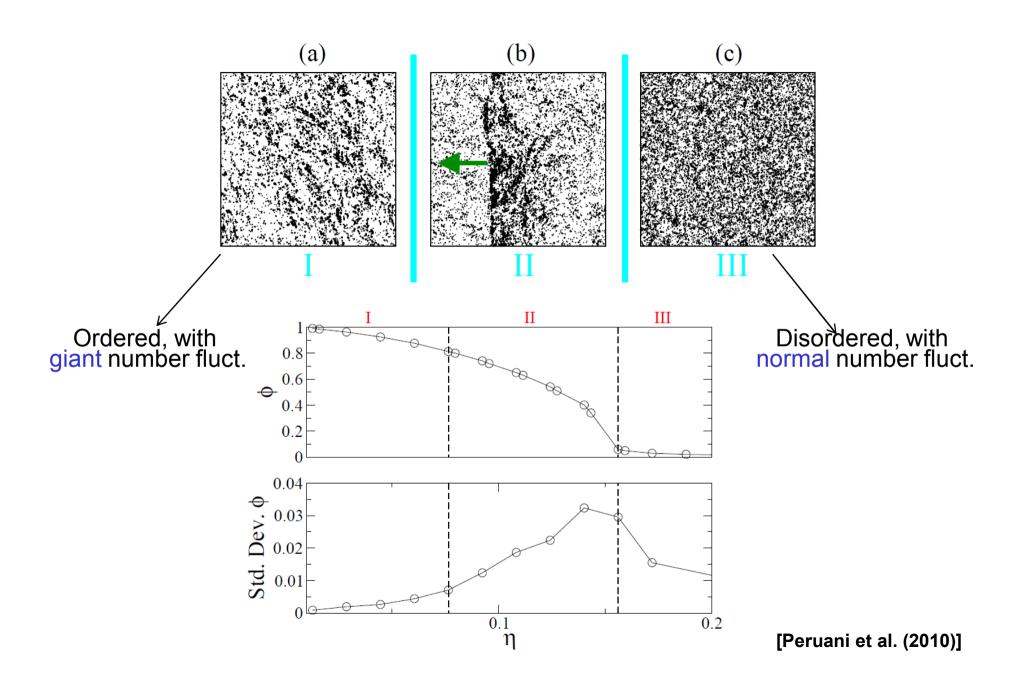
The phase transition and long-range order



[explain LRO, QLRO, and Short RO for finite size systems]

First vs second order transition





Number fluctuations

[at low cell densities]

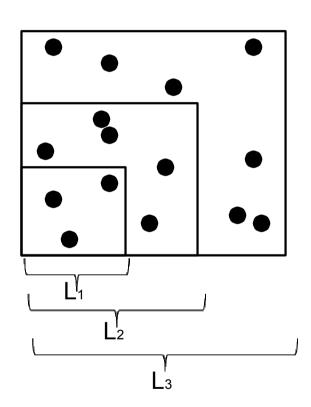
Average number:

$$\langle n(L) \rangle = \rho L^2$$

Average square number:

$$\Delta n^2 = <[n(L) - < n(L)>]^2>$$

$$\Delta n = (<[n(L) - < n(L)>]^2>)^{1/2} = < n(L)^{\frac{1}{2}}$$



n(L) = number of particles in box of size L

Normal number fluctuations

$$\Delta n = \langle n(L) \rangle^{1/2}$$

Number fluctuations

[at low cell densities]

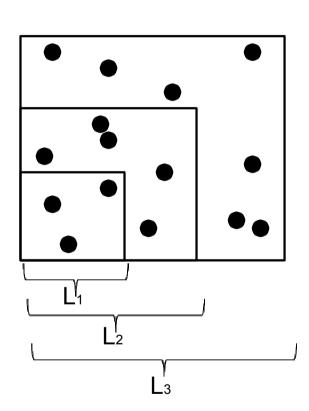
Average number:

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Average square number:

$$\Delta n^2 = <[n(L) - < n(L)>]^2>$$

$$\Delta n = (<[n(L) - < n(L)>]^2>)^{1/2} = < n(L)^{\beta}$$



n(L) = number of particles in box of size L

Giant number fluctuations

$$\Delta n = \langle n(L) \rangle^{\mu} \qquad \mu > 1/2$$

• A minimal continuum time SPP model with obstacles:

$$\dot{\mathbf{x}}_{i} = v_{0} \mathbf{V}(\theta_{i})$$

$$\dot{\theta}_{i} = g(\mathbf{x}_{i}) \left[\frac{\gamma_{b}}{n_{b}(\mathbf{x}_{i})} \sum_{|\mathbf{x}_{i} - \mathbf{x}_{j}| < R_{b}} \sin(\theta_{j} - \theta_{i}) \right]$$

$$+ h(\mathbf{x}_{i}) + \eta \xi_{i}(t),$$

$$h(\mathbf{x}_i) = \begin{cases} \frac{\gamma_o}{n_o(\mathbf{x}_i)} \sum_{|\mathbf{x}_i - \mathbf{y}_k| < R_o} \sin(\alpha_{k,i} - \theta_i) & \text{if } n_o(\mathbf{x}_i) > 0 \\ 0 & \text{if } n_o(\mathbf{x}_i) = 0 \end{cases},$$

A minimal continuum time SPP model with obstacles:

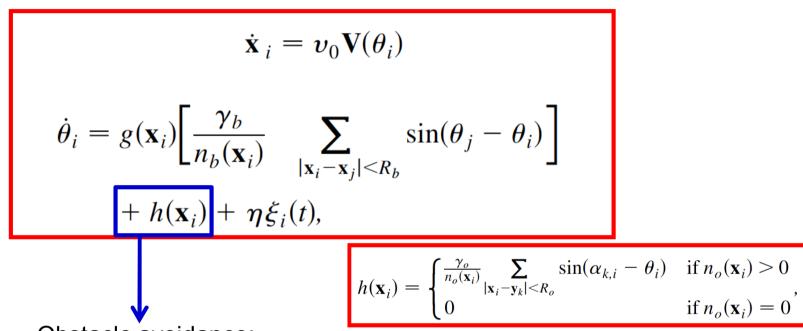
$$\dot{\mathbf{x}}_{i} = v_{0} \mathbf{V}(\theta_{i})$$

$$\dot{\theta}_{i} = g(\mathbf{x}_{i}) \left[\frac{\gamma_{b}}{n_{b}(\mathbf{x}_{i})} \sum_{|\mathbf{x}_{i} - \mathbf{x}_{i}| \leq R_{b}} \sin(\theta_{j} - \theta_{i}) \right] + h(\mathbf{x}_{i}) + \eta \xi_{i}(t),$$

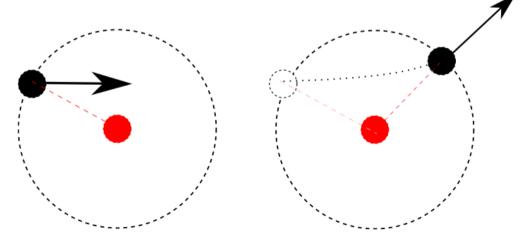
Continuum time version of Vicsek model (1995), i.e. a self-propelled XY model as introduced in Peruani, Deutch, Baer, Eur. J-ST (2008)

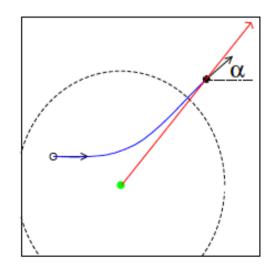
$$h(\mathbf{x}_i) = \begin{cases} \frac{\gamma_o}{n_o(\mathbf{x}_i)} \sum_{|\mathbf{x}_i - \mathbf{y}_k| < R_o} \sin(\alpha_{k,i} - \theta_i) & \text{if } n_o(\mathbf{x}_i) > 0 \\ 0 & \text{if } n_o(\mathbf{x}_i) = 0 \end{cases},$$

• A minimal continuum time SPP model with obstacles:



Obstacle avoidance:





• A minimal continuum time SPP model with obstacles:

$$\dot{\mathbf{x}}_{i} = v_{0} \mathbf{V}(\theta_{i})$$

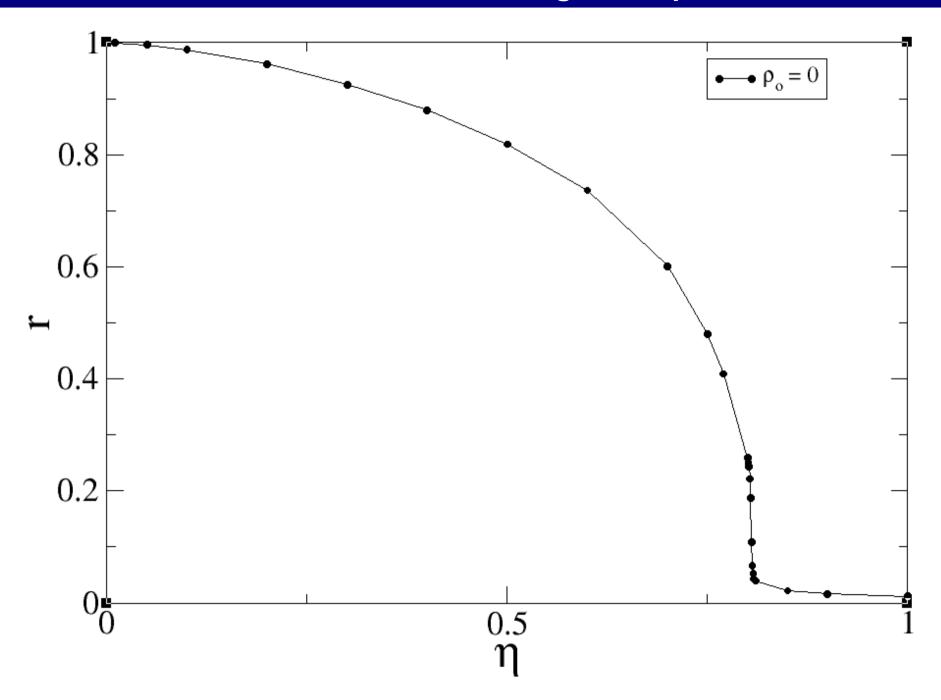
$$\dot{\theta}_{i} = g(\mathbf{x}_{i}) \left[\frac{\gamma_{b}}{n_{b}(\mathbf{x}_{i})} \sum_{|\mathbf{x}_{i} - \mathbf{x}_{j}| < R_{b}} \sin(\theta_{j} - \theta_{i}) \right]$$

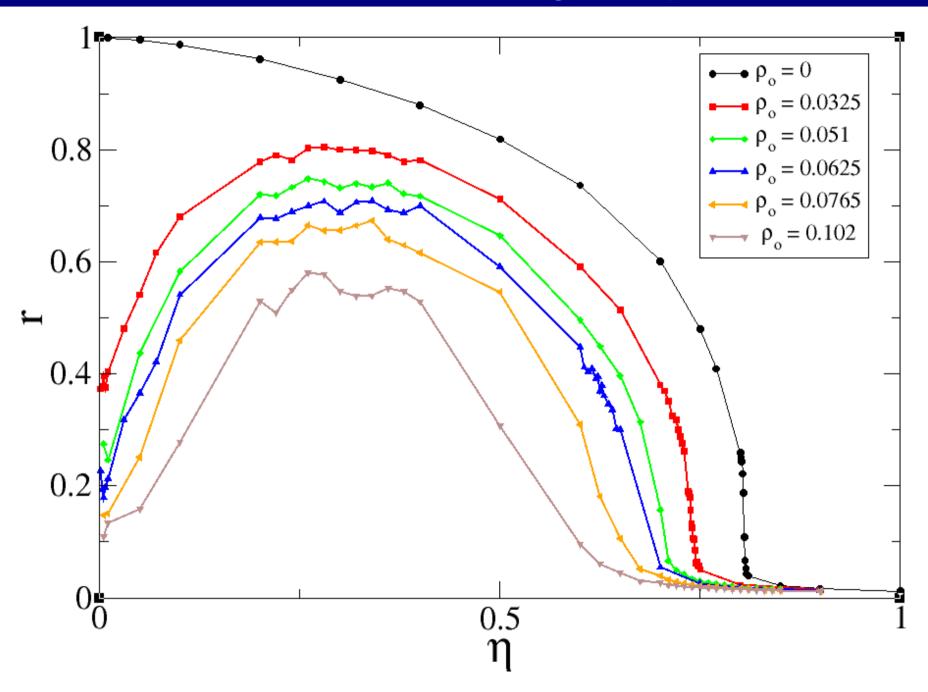
$$+ h(\mathbf{x}_{i}) + \eta \xi_{i}(t),$$

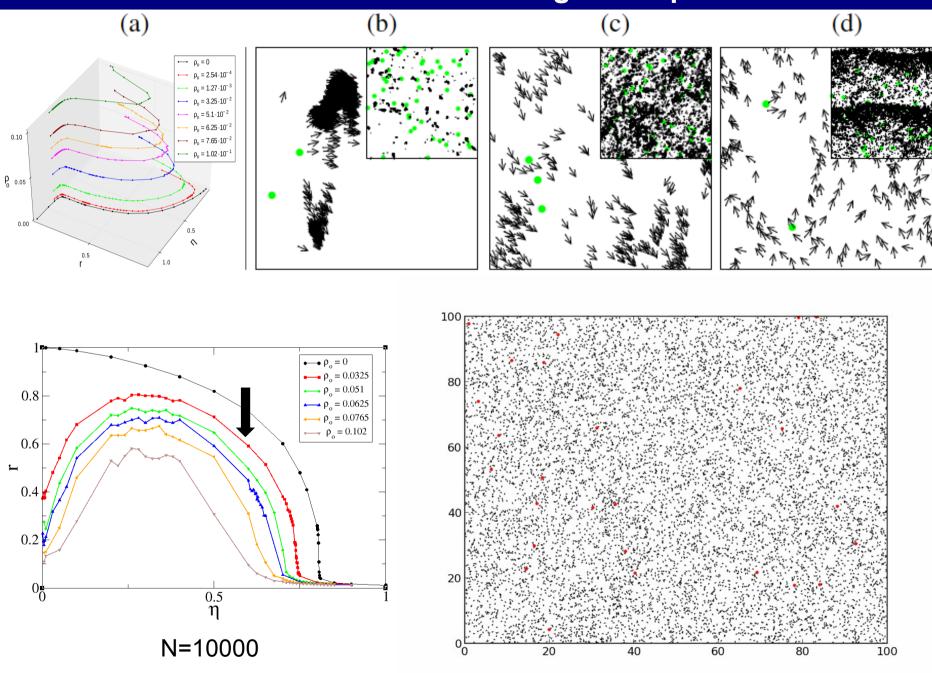
Order parameter
$$\longrightarrow$$
 $r=\langle r \rangle_t = \langle \left| \frac{1}{N_b} \sum_{i=1}^{N_b} e^{i\theta_i(t)} \right| \rangle_t$ [equiv. to the magnetization]

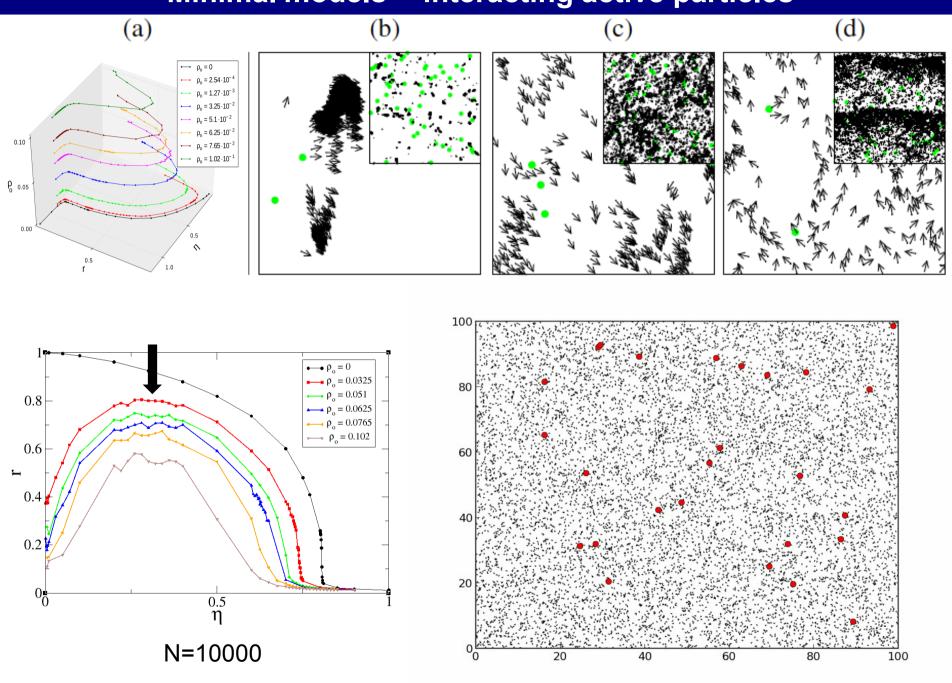
Susceptibility
$$\chi = \langle (r(t) - r)^2 \rangle_t$$

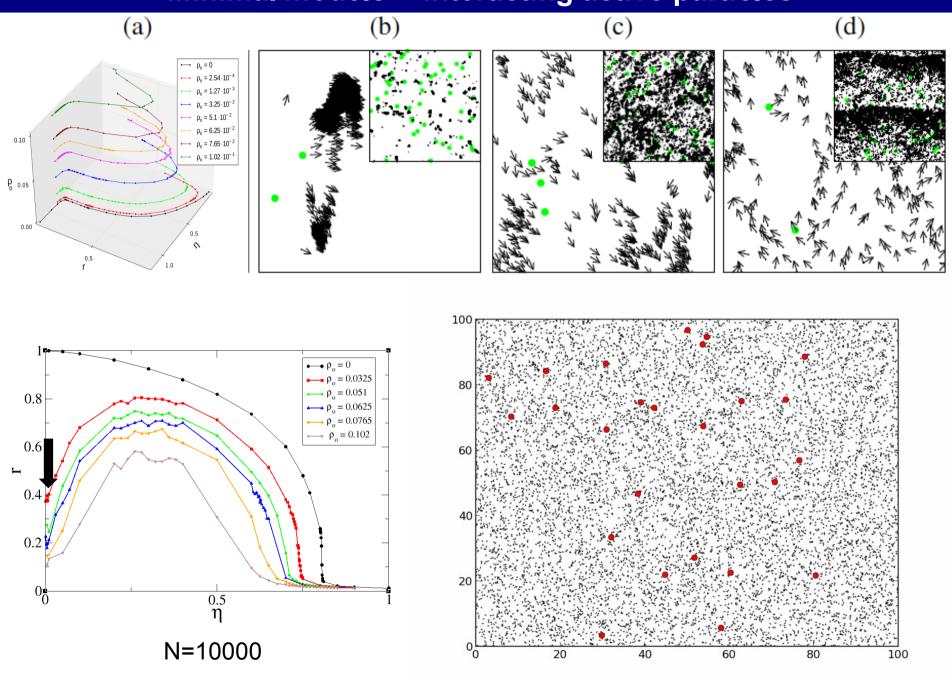
Binder cumulant
$$G = 1 - \frac{\langle r^4 \rangle_t}{3 \langle r^2 \rangle_t^2}$$



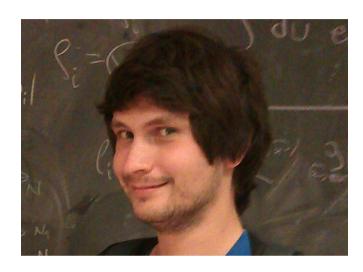








All this would have been impossible without the help of Sasha!



Oleksandr (Sasha) Chepizhko

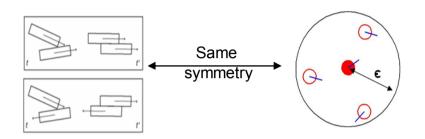
Chepizkho, Peruani, PRL (2013) Chepizkho, Altmann, Peruani, PRL (2013)

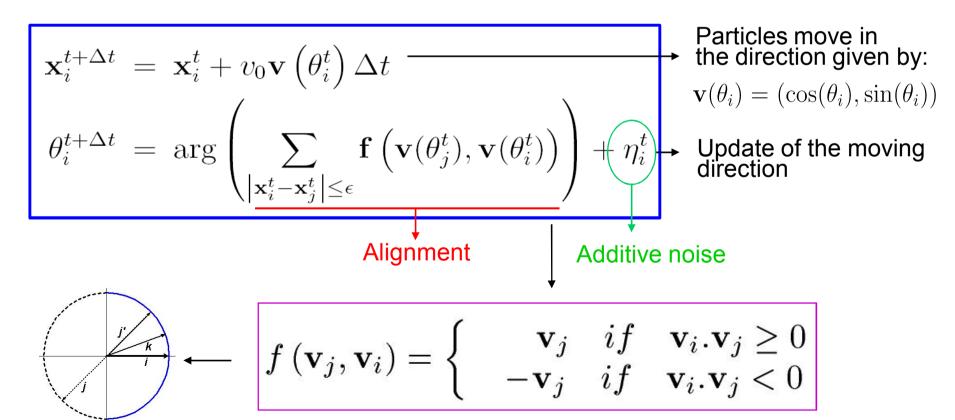
And many more to come in 2014!

Symmetries in active matter systems

A simple model for ("point-like") self-propelled rods (e.g., bacteria)

[F. Peruani, A. Deutsch, and M. Bär, Eur. Phys. J. Special Topics 157, 111 (2008)]





A simple model for ("point-like") self-propelled rods (e.g., bacteria)

[F. Peruani, A. Deutsch, and M. Bär, Eur. Phys. J. Special Topics 157, 111 (2008)]

$$\dot{\mathbf{x}}_i = v_0 e^{\mathrm{i}\,\theta_i}$$

$$\dot{\theta}_i = -\gamma \frac{\partial U}{\partial \theta_i}(\mathbf{x_i}, \theta_i) + \tilde{\eta}_i(t)$$

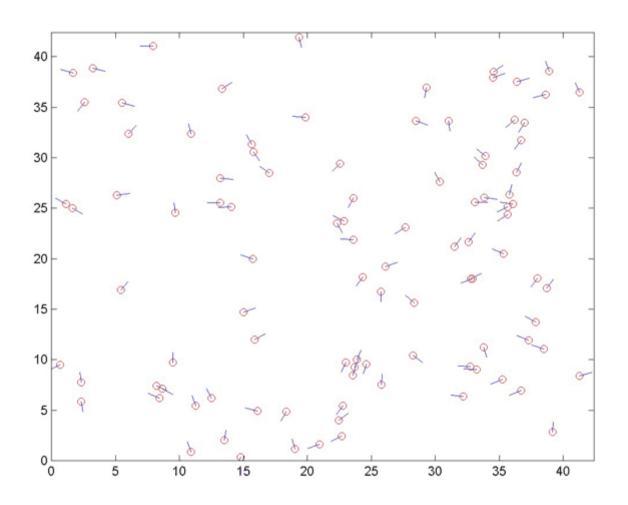
Ferromagnetic alignment
$$\rightarrow U(\theta, \theta') = -\cos(\theta - \theta')$$

Nematic alignment
$$\longrightarrow U(\theta, \theta') = -\cos^2(\theta - \theta')$$

Order parameters:

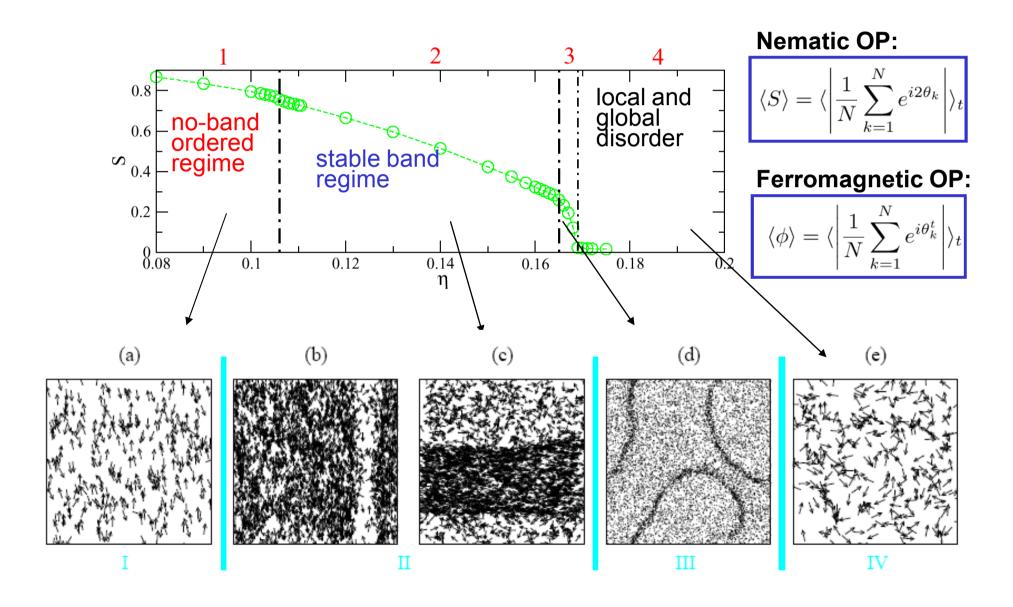
$$\phi = \left\langle \left| \frac{1}{N} \sum_{k=1}^{N} \exp(i \, \theta_k^t) \right| \right\rangle \qquad S = \left\langle \left| \frac{1}{N} \sum_{k=1}^{N} \exp(i \, 2 \, \theta_k^t) \right| \right\rangle$$

[Peruani, Deutsch, and Bär, EPJ ST 157, 111 (2008)]



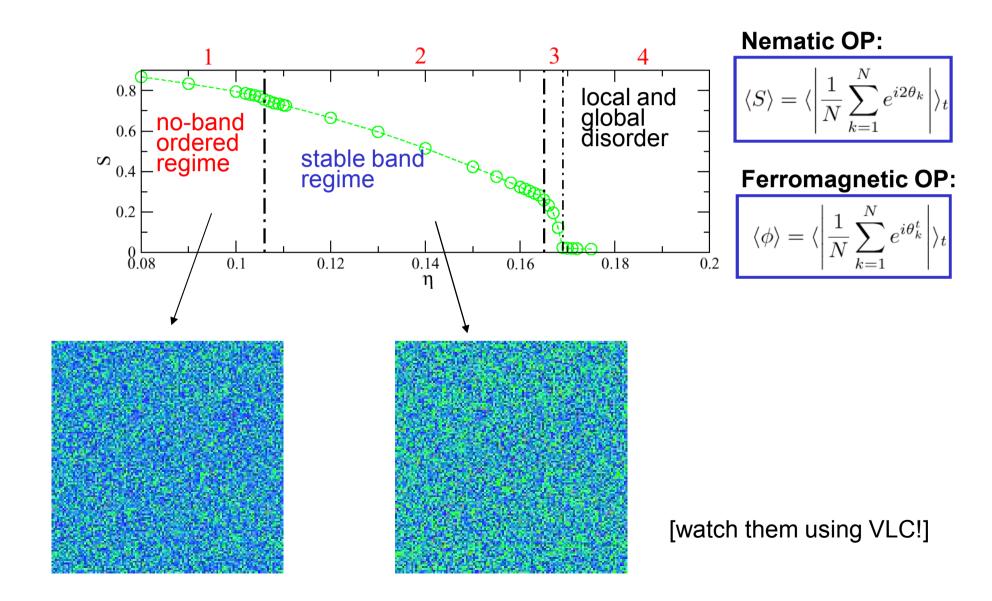
The symmetry of the alignment plays a crucial role in pattern formation

[F. Ginelli, F. Peruani, M. Bär, and H. Chaté, Phys. Rev. Lett. 104, 184502 (2010)]

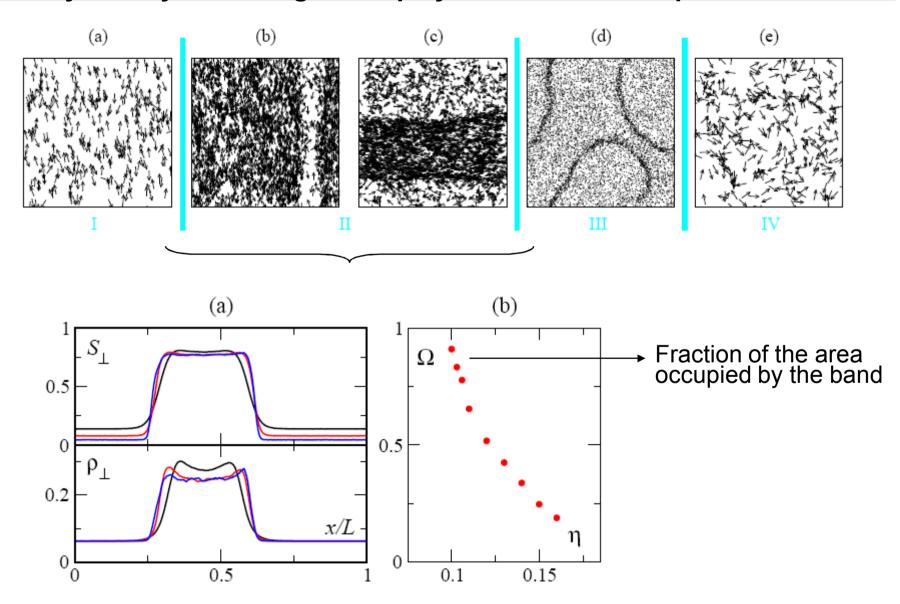


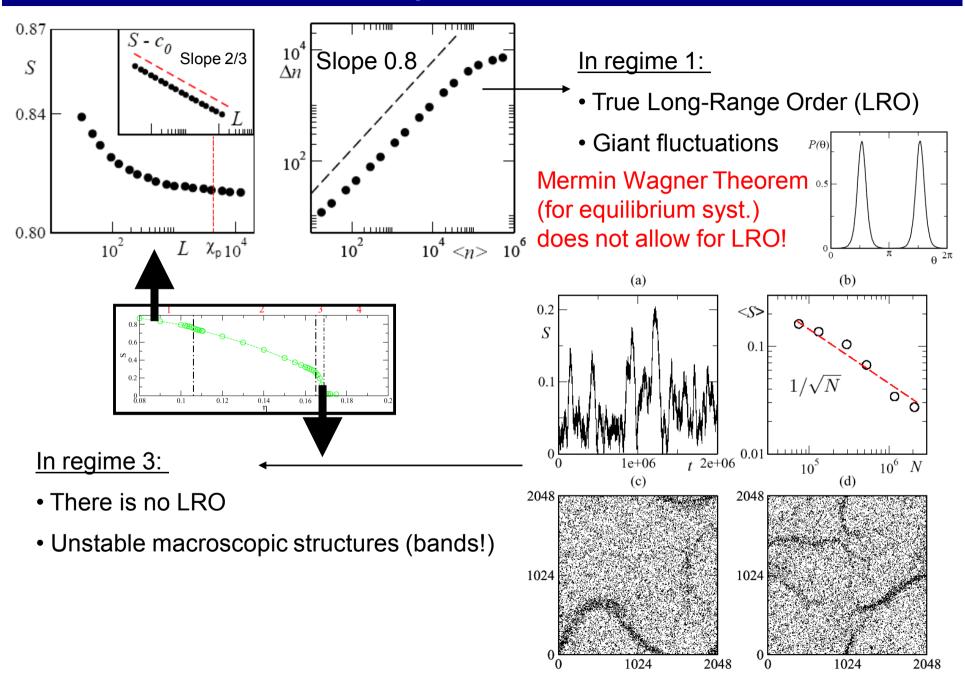
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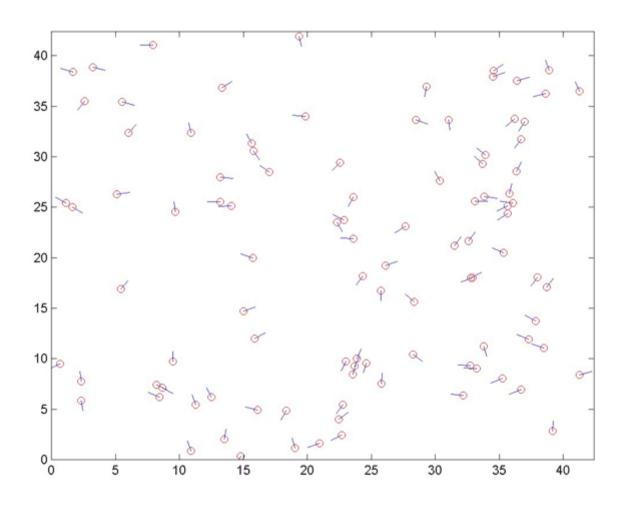


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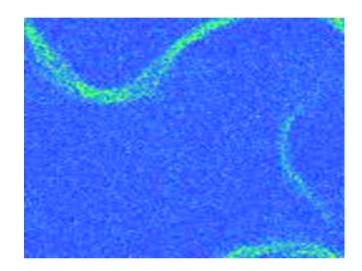


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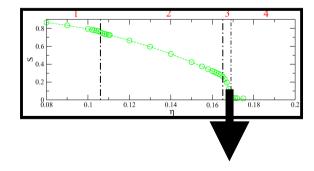


If we observe a very small system size or using a small box in a large system, particles behaves as shown in the movie

The symmetry of the alignment plays a crucial role in pattern formation



Green areas correspond to high density of particles, while blue means very low density



We are looking at the behavior of the system for this noise amplitude

The symmetry of the alignment determines the type of macroscopic order

• A mean-field approach to understand collective motion

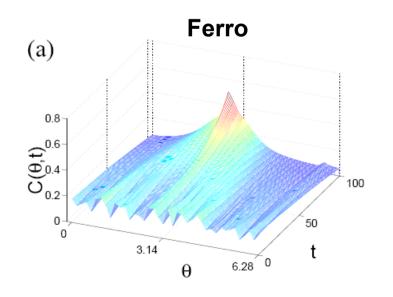
$$\partial_{t}C = D_{\theta}\partial_{\theta\theta}C + \gamma(\rho)\partial_{\theta}\left[\partial_{\theta}\left(\int_{0}^{2\pi}d\theta'C\left(\theta',t\right)U(\theta,\theta')\right)C\left(\theta,t\right)\right]$$

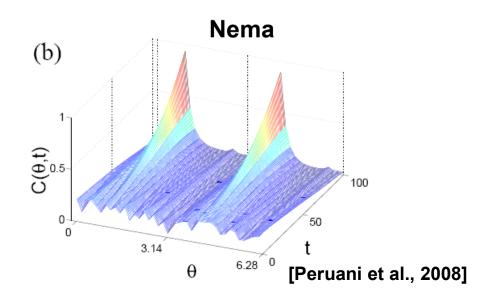
noise

alignment

Ferromagnetic alignment
$$\rightarrow U(\theta, \theta') = -\cos(\theta - \theta')$$

Nematic alignment
$$\longrightarrow U(\theta, \theta') = -\cos^2(\theta - \theta')$$





The symmetry of the alignment determines the type of macroscopic order

• A mean-field approach to understand collective motion

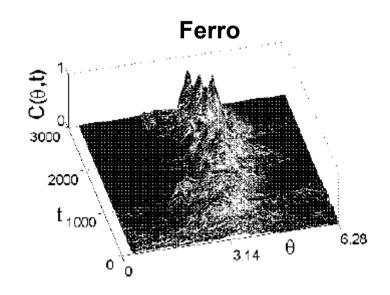
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noise

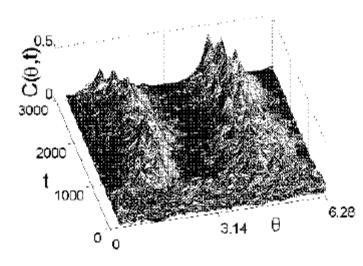
alignment

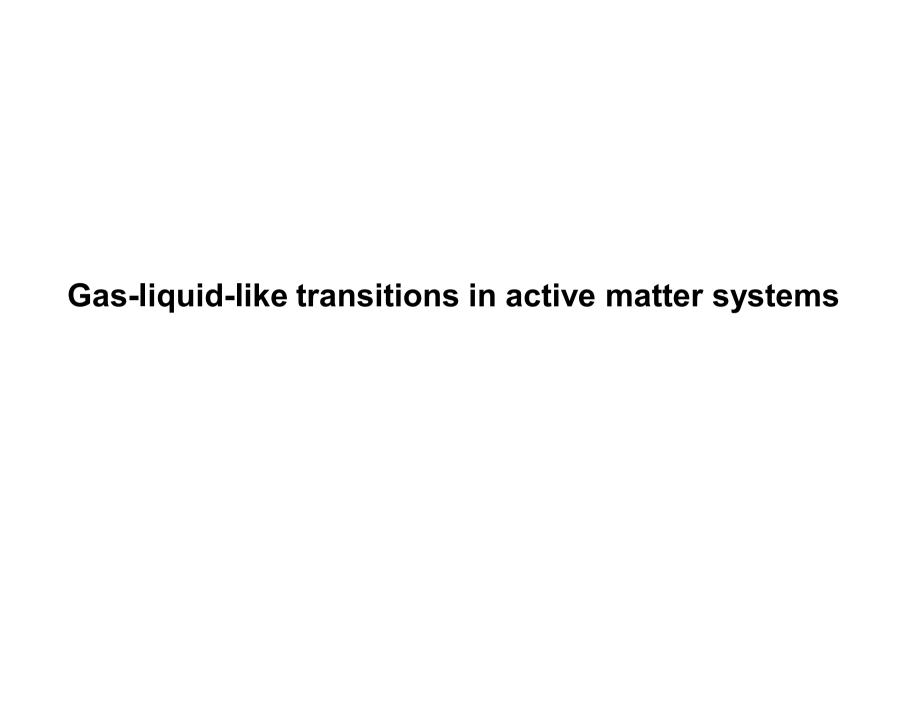
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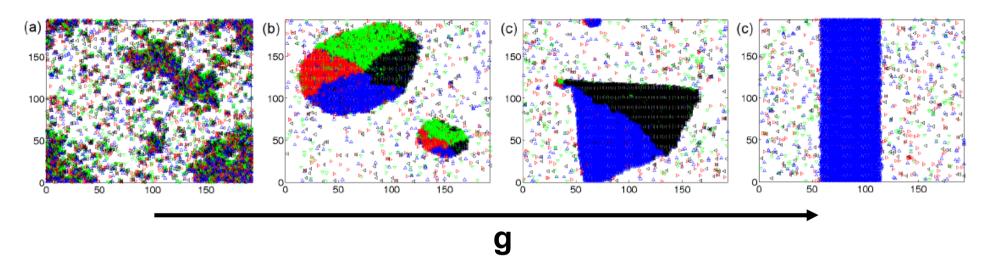


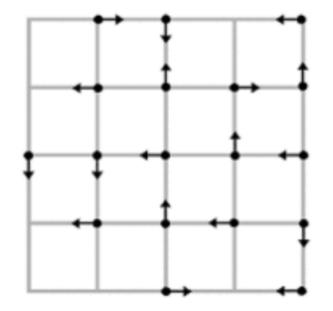
Nema





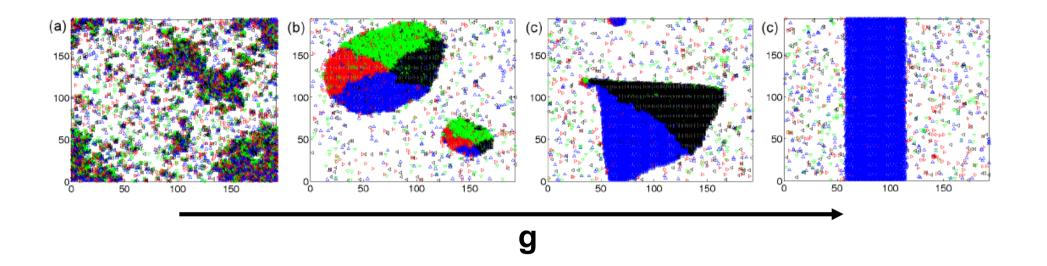
Coupling between local orientation, density and local particle speed!

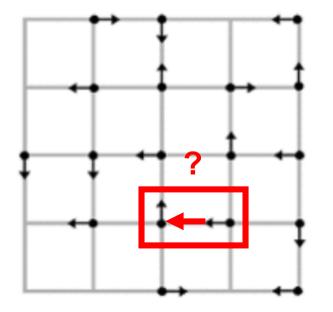




Each particle can perform two actions:

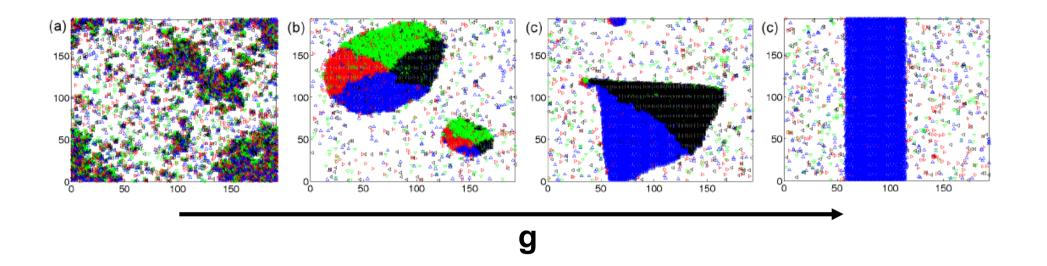
- 1) Migrate according to its velocity direction
- 2) Reorient its velocity direction

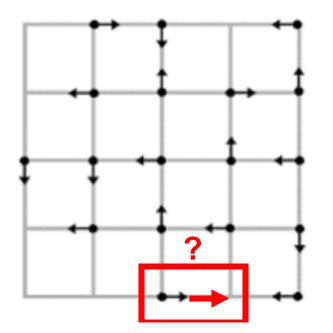




1) Migration according to its velocity direction

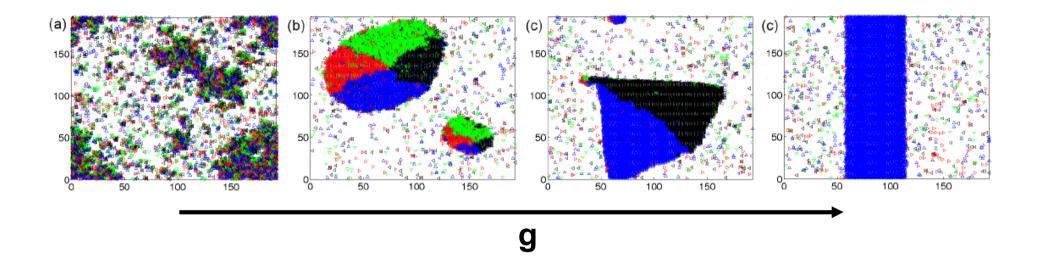
$$T_M((\mathbf{x}, \mathbf{v}) \to (\mathbf{y} = \mathbf{x} + \mathbf{v}, \mathbf{v})) = \begin{cases} m & \text{if node } \mathbf{y} \text{ is empty} \\ 0 & \text{if node } \mathbf{y} \text{ is occupied} \end{cases}$$

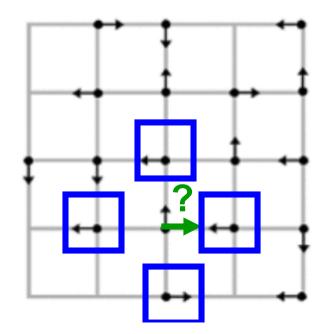




1) Migration according to its velocity direction

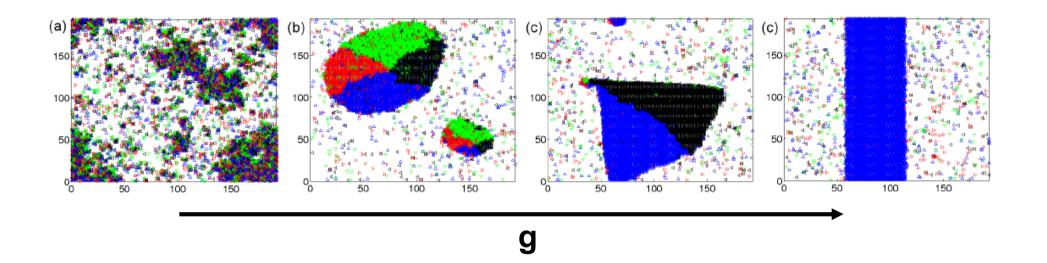
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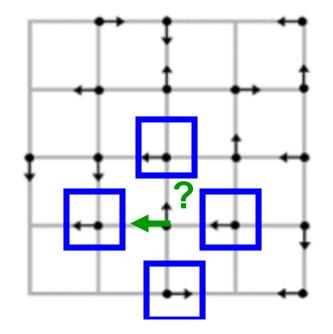




2) Reorient its velocity direction

$$T_R((\mathbf{x}, \mathbf{v}) \to (\mathbf{x}, \mathbf{w})) = \exp(g \sum_{\mathbf{y} \in A(\mathbf{x})} \langle \mathbf{w} | \mathbf{V}(\mathbf{y}) \rangle)$$

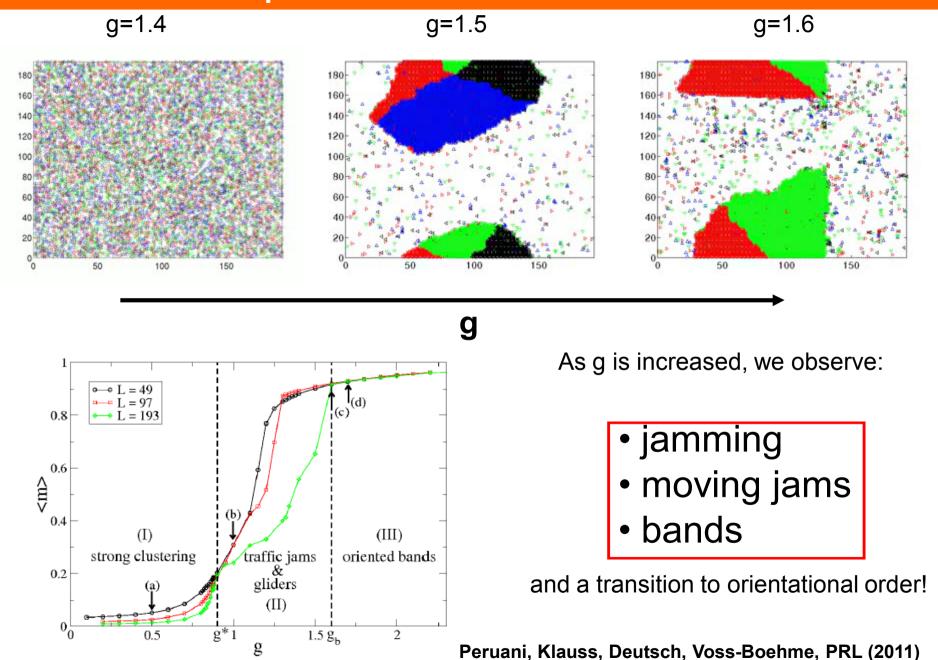


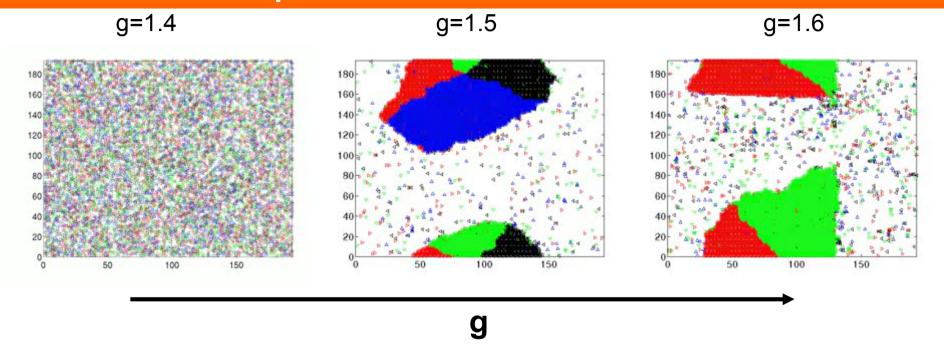


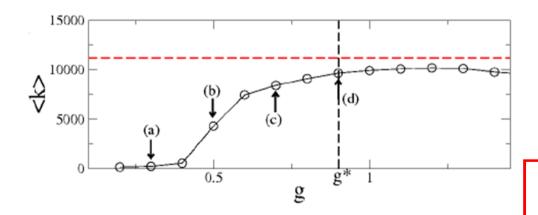
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Peruani, Klauss, Deutsch, Voss-Boehme, PRL (2011)







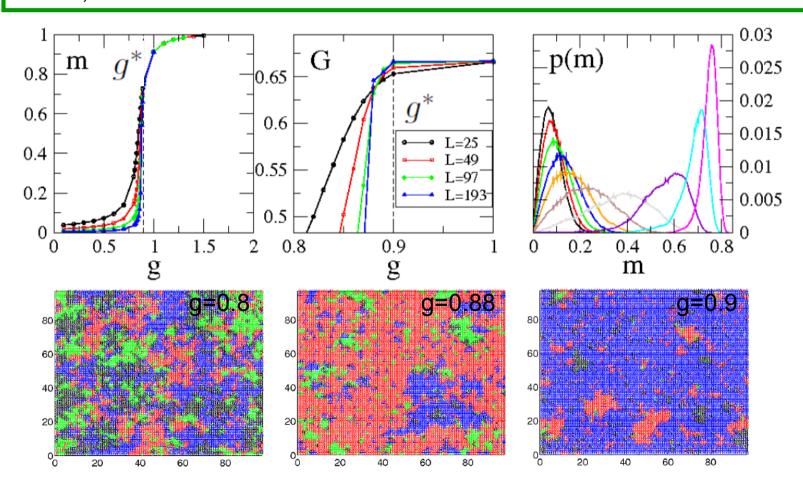
- Gas-liquid-like transition
- Second orientational order

Different behavior at low and high densities (particularly evident at full occupancy)

Order of the phase transition at high and low density

Results at "high" (=full occupancy) density

- The problem with full occupancy can be mapped onto the 4-Potts model
- The 4-Potts model exhibits a second order phase transition
- Then, our model exhibits a **second order** transition at **d=1**!



Are these transition similar to the equilibrium gas-liquid transitions?

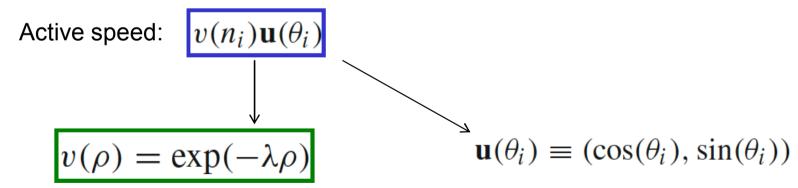
Can we map non-equilibrium to equilibrium?

Equations of motion for the i-th particle:

$$\dot{\mathbf{x}}_i = v(n_i)\mathbf{u}(\theta_i) + \sqrt{2D_x}\boldsymbol{\sigma}_i(t)$$

$$\dot{\theta}_i = -\frac{\gamma}{n_i} \sum_{j=1}^N g(|\mathbf{x}_i - \mathbf{x}_j|/R) \sin(\theta_i - \theta_j) + \sqrt{2D_\theta}\eta_i(t)$$

Number of neighbors: $n_i = \sum_{j=1}^N g(|\mathbf{x}_i - \mathbf{x}_j|/R)$



dependency with the number of neighbors/local density

We can write the previous equations in the following dimensionless form:

$$\frac{d\tilde{\mathbf{x}}_{i}}{d\tilde{t}} = \tilde{\mathbf{v}}_{0}(n_{i})\mathbf{u}(\theta_{i}) + \epsilon\sqrt{2\tilde{D}_{x}}\boldsymbol{\sigma}_{i}(\tilde{t})$$

$$\frac{d\theta_{i}}{d\tilde{t}} = -\frac{\bar{\gamma}}{n_{i}}\sum_{j=1}^{N}g((\tilde{\mathbf{x}}_{i} - \tilde{\mathbf{x}}_{j})/\alpha)\sin(\theta_{i} - \theta_{j}) + \sqrt{2}\eta_{i}(\tilde{t})$$

Evolution of the empirical density

$$\frac{\partial f_d}{\partial t} = -\varepsilon \nabla \cdot (v(\rho(\mathbf{x}, t))) \mathbf{u}(\theta) f_d(\mathbf{x}, \theta, t)
+ \frac{\bar{\gamma}}{\rho(\mathbf{x}, t)} \frac{\partial}{\partial \theta} \left(f_d(\mathbf{x}, \theta, t) \int d\theta' \sin(\theta - \theta') f_d(\mathbf{x}, \theta', t) \right)
+ \frac{\partial^2 f_d}{\partial \theta^2} + \varepsilon^2 D_x \nabla^2 f_d + \sqrt{\frac{2}{N}} \frac{\partial}{\partial \theta} \left(\eta(\mathbf{x}, \theta, t) \sqrt{f_d} \right)
+ \varepsilon \frac{\sqrt{2D_x}}{\sqrt{N}} \nabla \cdot \left(\sigma(\mathbf{x}, \theta, t) \sqrt{f_d} \right),$$

Gaussian noise delta correlated

Instead of working with fd with can look at the evolution of the first n Fourier modes of fd:

$$\rho(\mathbf{x},t) = \int d\theta f_d(\mathbf{x},\theta,t) \qquad \longrightarrow \text{ density}$$

$$\mathbf{P}(\mathbf{x},t) = \int d\theta \mathbf{u}(\theta) f_d(x,y,\theta,t) \qquad \longrightarrow \text{ polarization}$$

Their evolution are given by:

$$\frac{\partial \rho}{\partial t} = -\varepsilon \nabla \cdot (v \mathbf{P}) + \varepsilon^2 D_x \nabla^2 \rho + \varepsilon \frac{\sqrt{2D_x}}{\sqrt{N}} \nabla \cdot (\mathbf{\xi}(x, y, t))$$

$$\frac{\partial \mathbf{P}}{\partial t} = -\frac{1}{2} \varepsilon \nabla (v \rho) + \left(\frac{\bar{\gamma}}{2} - 1\right) \mathbf{P} + \varepsilon^2 D_x \nabla^2 \mathbf{P} + \sqrt{\frac{2}{N}} \boldsymbol{\eta}(x, y, t) + O\left(\frac{\varepsilon}{\sqrt{N}}\right)$$

The noise term are given by:

$$\eta_{x}(x, y, t) = \int d\theta \sin \theta \sqrt{f_{d}} \eta(x, y, \theta, t)$$

$$\eta_{y}(x, y, t) = -\int d\theta \cos \theta \sqrt{f_{d}} \eta(x, y, \theta, t)$$

$$\xi(x, y, t) = \int d\theta \sqrt{f_{d}} \sigma(x, y, \theta, t).$$

With correlations:

$$\langle \eta_{x}(x, y, t) \eta_{x}(x', y', t') \rangle \simeq \delta(x - x') \delta(y - y') \delta(t - t') \frac{1}{2} \rho(x, y, t)$$

$$\langle \eta_{y}(x, y, t) \eta_{y}(x', y', t') \rangle \simeq \delta(x - x') \delta(y - y') \delta(t - t') \frac{1}{2} \rho(x, y, t)$$

$$\langle \eta_{x}(x, y, t) \eta_{y}(x', y', t') \rangle \simeq 0$$

$$\langle \xi(x, y, t) \xi(x', y', t') \rangle = \delta(x - x') \delta(y - y') \delta(t - t') \rho(x, y, t).$$

The evolution of P is faster than the evolution of the density and asymptotically we expect P=0, which means that at order ϵ we can expect:

$$\mathbf{P} = \varepsilon \frac{-1}{2\left(1 - \frac{\bar{\gamma}}{2}\right)} \nabla(v(\rho)\rho) + \sqrt{\frac{2}{N}} \frac{1}{1 - \frac{\bar{\gamma}}{2}} \boldsymbol{\eta}$$

If we insert this expression into the equation for the density we find:

$$\frac{\partial \rho}{\partial t} = \frac{1}{2} \nabla \cdot \left(\frac{v}{1 - \frac{\bar{\gamma}}{2}} \nabla [v(\rho)\rho] \right) + D_x \nabla^2 \rho$$
$$+ \frac{\sqrt{2D_x}}{\sqrt{N}} \nabla \cdot (\xi(\mathbf{r}, t)) + \sqrt{\frac{2}{N}} \nabla \cdot \left(\frac{v}{1 - \frac{\bar{\gamma}}{2}} \eta \right)$$

The previous equation can be rewriten as a Lagevin eq. of the following form:

$$\frac{\partial \rho}{\partial t} = U[\rho](\mathbf{x}) + \frac{1}{\sqrt{N}} \nu(\mathbf{x}, t)$$

Where we define:

$$U[\rho](\mathbf{x}) = \frac{1}{2} \nabla \cdot \left(\frac{v(\rho)}{1 - \frac{\bar{\gamma}}{2}} \nabla [v(\rho)\rho] \right) + D_x \nabla^2 \rho$$

$$\langle v(x, y, t)v(x', y', t')\rangle = D[\rho](\mathbf{x}, \mathbf{x}')\delta(t - t')$$

and in addition:

$$b[\rho] = 2D_x \rho + \frac{\rho v^2(\rho)}{\left(1 - \frac{\bar{\gamma}}{2}\right)^2}$$

$$D[\rho](\mathbf{x}, \mathbf{x}') = \partial_x \partial_{x'} [b[\rho](\mathbf{x}) \delta(\mathbf{x} - \mathbf{x}')] + \partial_y \partial_{y'} [b[\rho](\mathbf{x}) \delta(\mathbf{x} - \mathbf{x}')]$$

We derive the associated FP for the previous Lagevin equation:

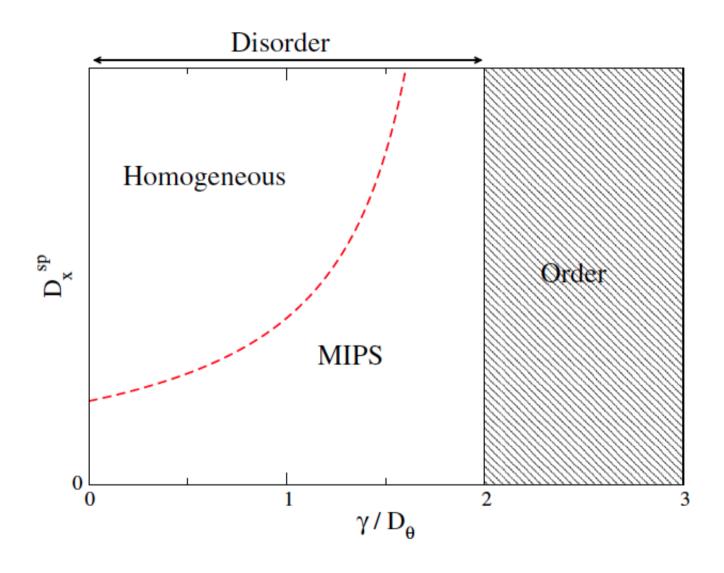
$$\frac{\partial \mu_t}{\partial t} = -\int d\mathbf{x} \frac{\delta}{\delta \rho(\mathbf{x})} \left(U[\rho](\mathbf{x}) \mu_t \right)
+ \frac{1}{2N} \int d\mathbf{x} \frac{\delta}{\delta \rho(\mathbf{x})} \left\{ \int d\mathbf{x}' D[\rho](\mathbf{x}, \mathbf{x}') \frac{\delta}{\delta \rho(\mathbf{x}')} \mu_t \right\}$$

()We look for solution of the following type (formal solution of the previous eq.):

$$\mu [\rho] \sim e^{NS[\rho]} \longrightarrow S[\rho] = \int d\mathbf{x} s(\rho(\mathbf{x}))$$

After some calculations, we arrive at:

$$s''(\rho) = -\left(\frac{v^2(\rho) + \rho v(\rho)v'(\rho)}{\left(1 - \frac{\bar{\gamma}}{2}\right)b[\rho]} + \frac{2D_x}{b[\rho]}\right)$$



Summary

- 1. Definition of active soft-matter
- 2. Examples in biology and non-living systems
- 3. Minimal models of active particles
- 4. Symmetries in active particle systems
- 5. Gas-liquid-like transitions

