

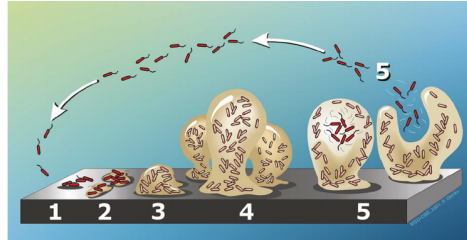
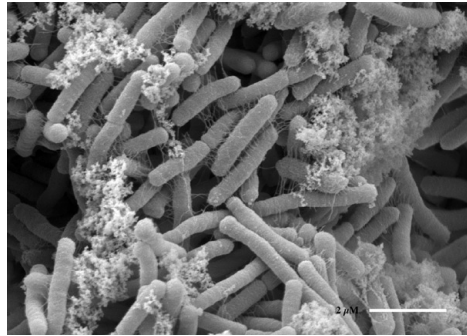
Non-equilibrium self-organization of motile bacteria with fluctuating population and speed

Tobias Grafke, M. Cates, E. Vanden-Eijnden

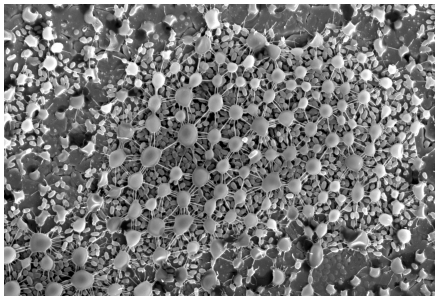
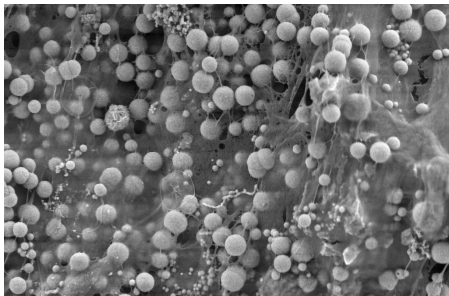
Collective behavior of bacteria

Bacteria show complex collective behavior

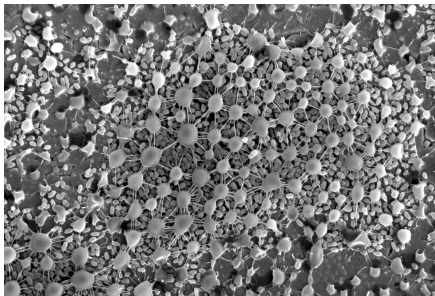
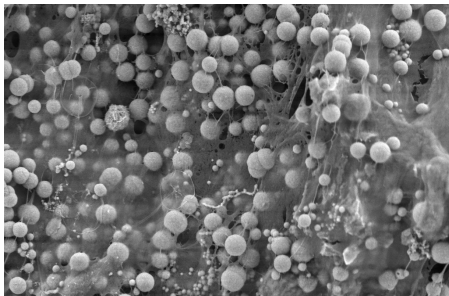
- For example, bacteria such as *E. Coli* are capable of **active propulsion**, i.e. have a free-swimming (*planktonic*) stage.
- They are capable of sensing their environment through **quorum sensing**, density-dependent gene regulation.
- They stick to surface to form **biofilms**, high density colonies.
- They exhibit **cyclic**/time-periodic behavior: Biofilm formation, maturation, dispersion, planktonic stage.
- All these are controlled by highly complex bio-chemical processes.



Collective behavior of bacteria



Collective behavior of bacteria



Statistical mechanics of active matter

Stepping away from the biological complexity: Is it possible to describe similarly complex life-cycles as **emergent behavior** of a large number of simple **individual agents** subject to a small number of collective rules?

Intrinsically **out-of-equilibrium** system: Statistical mechanics of **active matter**

Continuum description of motile bacteria

- Self-propelled bacteria modeled as **active Brownian motion**:

For position $X \in \Omega \subset \mathbb{R}^d$, direction $\hat{n} \in S^{d-1}$, and location dependent swim speed $v(X) \in \mathbb{R}$

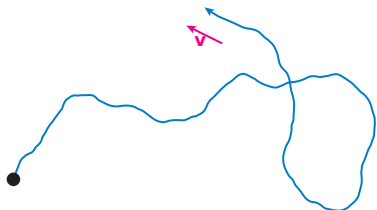
$$\dot{X} = v(X)\hat{n},$$

$$d\hat{n} = \tau^{-1/2}P \circ dW, \quad P = \text{Id} - \hat{n}\hat{n}^T$$

Direction diffuses on S^{d-1} with tumbling rate τ^{-1} .

- Limit $\tau \rightarrow 0$ yields Brownian motion

$$dX = \sqrt{2D(X)} \circ dW \quad \text{with diffusivity} \quad D(x) = \tau v^2(x)$$



active Brownian motion[†]

[†] M. Cates, J. Tailleur (2015)

Continuum description of motile bacteria

Cates & Tailleur

Now consider N such particles with position $X_i, i \in \{1, \dots, N\}$.

To model **quorum sensing**, introduce scale δ over which particles feel each other's influence,

$$dX_i = \sqrt{2D(\rho_{N,\delta}(t, X_i))} \circ dW_i$$

with

$$\rho_{N,\delta}(t, x) = \int_{\Omega} \phi_{\delta}(x - y) \rho_N(t, y) dy, \quad \rho_N(t, x) = \frac{1}{N} \sum_{j=1}^N \delta(x - X_j(t))$$

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In the limit $N \rightarrow \infty$, yields closed integro-differential equation for $\rho_N \rightarrow \rho$,

$$\partial_t \rho = \nabla \cdot (D(\rho_{\delta}) \nabla \rho + \frac{1}{2} D'(\rho_{\delta}) \nabla \rho_{\delta}), \quad \rho_{\delta}(t, x) = \int_{\Omega} \phi_{\delta}(x - y) \rho(t, y) dy$$

in the sense that

$$\forall \epsilon, T > 0 : \lim_{N \rightarrow \infty} \mathbb{P} \left(\sup_{0 \leq t \leq T} \left| \frac{1}{N} \sum_{j=1}^N f(X_j(t)) - \int_{\Omega} \rho(t, x) f(x) dx \right| > \epsilon \right) = 0$$

Continuum description of motile bacteria

To make this closed in $\rho(t, x)$, consider $D(\rho) = D_0 e^{-\rho}$, and expand in $\delta \ll 1$,

$$\rho_\delta(x) \approx \rho(x) + \frac{1}{2} \delta^2 \partial_x^2 \rho(x).$$

Then we obtain an **effective diffusion** equation

$$\partial_t \rho = \nabla \cdot (D_e(\rho) \nabla \rho - \delta^2 \rho D(\rho) \nabla \Delta \rho)$$

with diffusivity

$$D_e(\rho) = D(\rho) + \frac{1}{2} D'(\rho) \rho.$$

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Despite non-equilibrium microscopic model, continuum model restores **detailed balance**

$$\partial_t \rho = \nabla \cdot (\rho D(\rho) \nabla (\delta E / \delta \rho))$$

with

$$E(\rho) = \int_{\Omega} (\rho \log \rho - \rho + f(\rho) + \frac{1}{2} \delta^2 |\nabla \rho|^2) dx, \quad f'(\rho) = \frac{1}{2} \log D(\rho) \quad (\text{free energy})$$

(“Thermodynamic mapping” depends on regularization, not preserved by original Cates & Tailleur)

Motility induced phase separation

$$\partial_t \rho = \nabla \cdot (D_e(\rho) \nabla \rho - \delta^2 \rho D(\rho) \nabla \Delta \rho),$$

$$D_e(\rho) = D(\rho) + \frac{1}{2} D'(\rho) \rho$$

$D(\rho) > 0$ **diffusivity**, but $D_e(\rho) < 0$ possible!

For $D(\rho) = D_0 e^{-\rho}$

- effective diffusivity

$$D_e(\rho) = D_0 \left(1 - \frac{1}{2} \rho\right) e^{-\rho}$$

- Homogeneous configuration stable if

$$\bar{\rho} = \frac{1}{|\Omega|} \int_{\Omega} \rho(x) dx < 2$$

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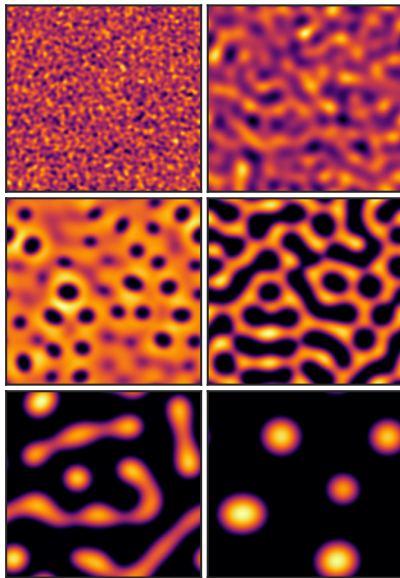
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- If $\bar{\rho} > 2$, then **phase separation** occurs

Feedback loop: Accumulation induced slowdown + slowdown induced accumulation



Motility induced phase separation

Up to now, single behavioral rule:

Active Brownian motion with **density dependent diffusivity**

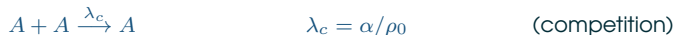
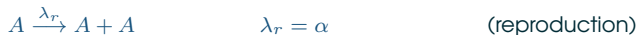
Introduce second behavioral rule: **Population dynamics**

Motility induced phase separation

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Active Brownian motion with **density dependent diffusivity**

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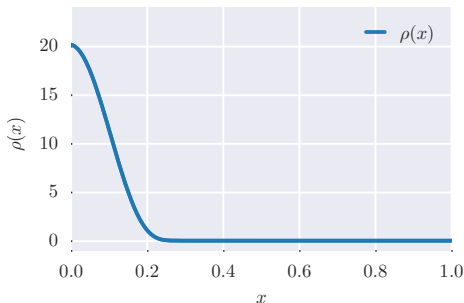
with **carrying capacity** ρ_0 and **timescale** α , leads to **logistic growth**.

$$\partial_t \rho = \nabla \cdot (D_e(\rho) \nabla \rho - \delta^2 \rho D(\rho) \nabla \Delta \rho) + \alpha \rho (1 - \rho/\rho_0)$$

- Phase separation eventually if

$$D_e(\rho_0) < 0$$

- System will drive itself into instability
- Coarsening stopped by growth, **arrested phase separation**[†]



[†] Cates, M. E. and Marenduzzo, D. and Pagonabarraga, I. and Tailleur, J. (2010)

Timescale separation

Consider large **timescale separation**, $\alpha \ll 1$, i.e. fast propulsion, slow reproduction

$$\partial_t \rho = \nabla \cdot (D_e(\rho) \nabla \rho - \delta^2 \rho D(\rho) \nabla \Delta \rho) + \alpha \rho (1 - \rho/\rho_0)$$

Asymptotic analysis:

- Propulsion is **conservative**,
 $\bar{\rho}$ is invariant under fast dynamics
- Emergence of **slow manifold** \mathcal{M} ,

$$0 = \nabla \cdot (D_e(\rho) \nabla \rho - \delta^2 \rho D(\rho) \nabla \Delta \rho)$$

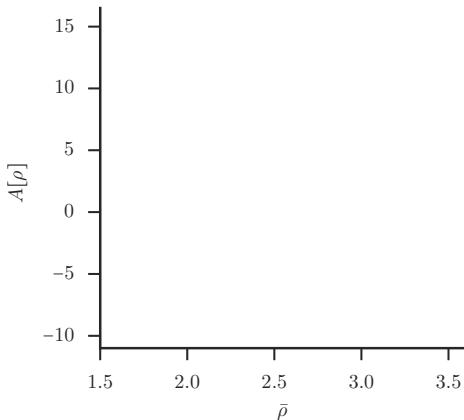
foliated by $\bar{\rho}$.

- Obtain **reduced dynamics**

$$\dot{\bar{\rho}} = \alpha(\bar{\rho} - \bar{\rho}^2/\rho_0), \quad \rho = \rho_{\mathcal{M}}(x, \bar{\rho}(t))$$

- 3 regimes

$\rho_0 < \rho_S$	homogeneous solution
$\rho_S < \rho_0 < \rho_c$	limit cycles
$\rho_c < \rho_0$	two stable fixed points



$$A[\rho] = \int_0^{1/2} \rho(x) dx - \int_{1/2}^1 \rho(x) dx$$

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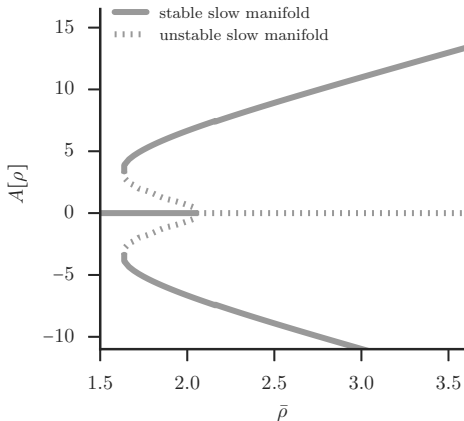
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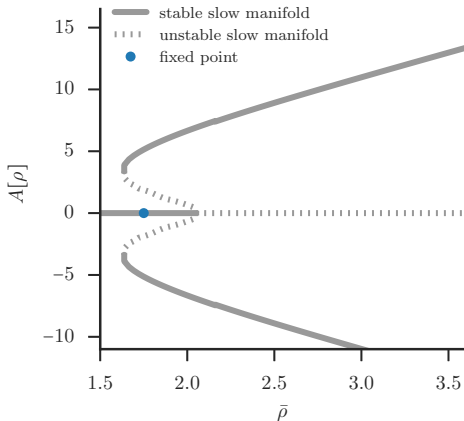
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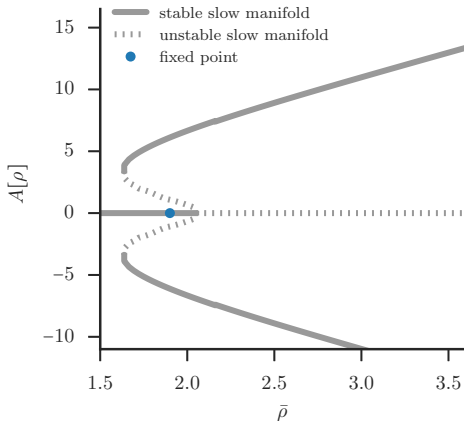
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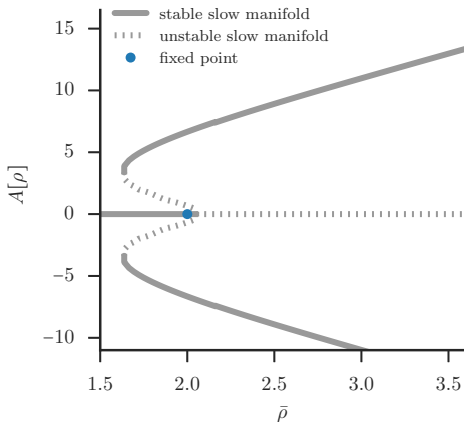
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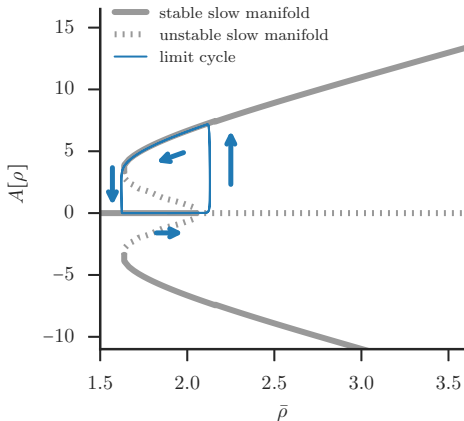
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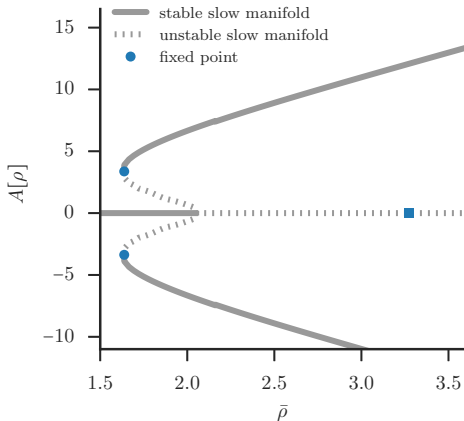
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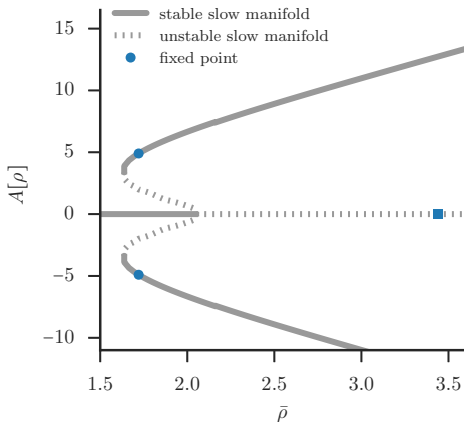
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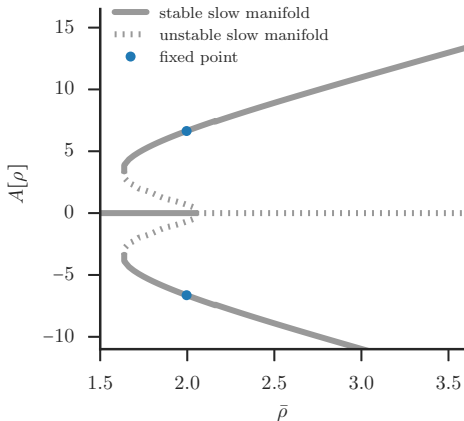
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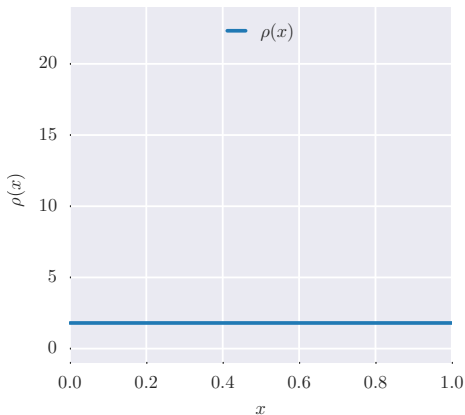
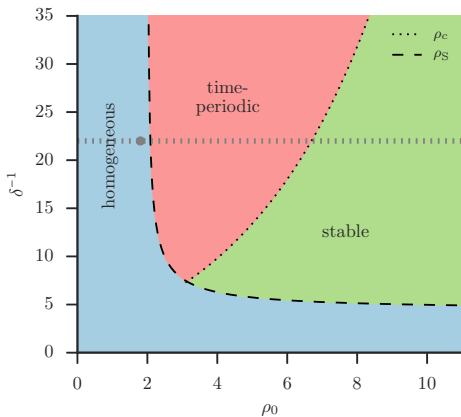
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Homogeneous regime:

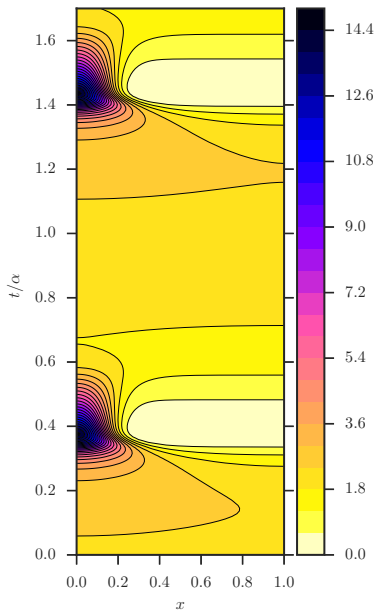
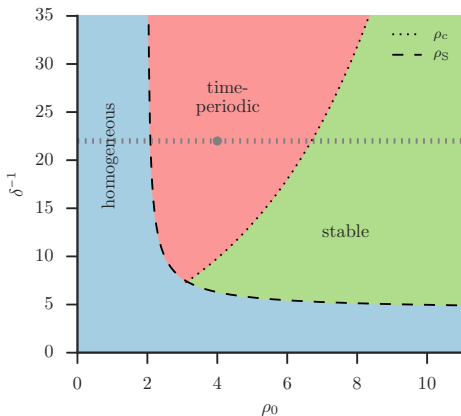
$$\rho_0 < \rho_c$$

$$\Omega = [0, 1]$$

Neumann boundary conditions,

$$\partial_x \rho(0) = \partial_x \rho(1) = 0$$

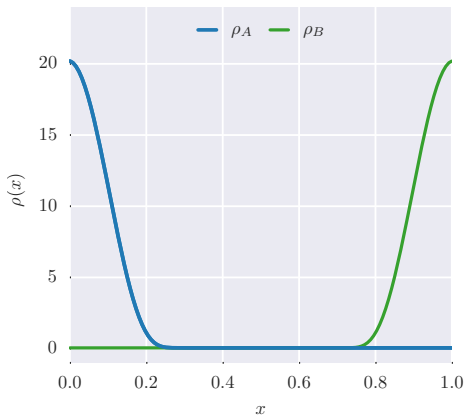
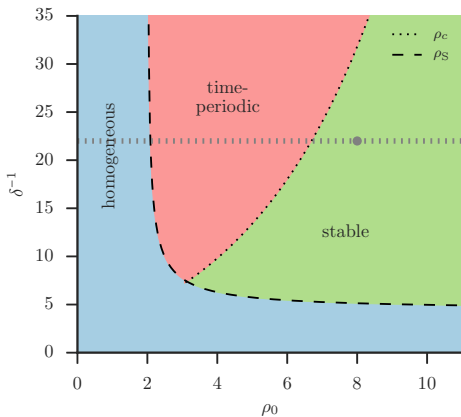
Timescale separation



Time periodic regime:

$$\rho_c < \rho_0 < \rho_S$$

Timescale separation



Stable regime:

$$\rho_c < \rho_0$$

$$\Omega = [0, 1]$$

Neumann boundary conditions,

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The effect of fluctuations

$$\partial_t \rho = \nabla \cdot (D_e(\rho) \nabla \rho - \delta^2 \rho D(\rho) \nabla \Delta \rho) + \alpha \rho (1 - \rho / \rho_0)$$

- Is a **law of large numbers** (LLN) for $N \rightarrow \infty$
(where N typical number of particles)
- Gaussian fluctuations around these dynamics captured by **central limit theorem** (CLT)
- We are interested in long time behavior: **Large deviation theory** (LDT)
- Both **propulsion** and **reproduction** are subject to fluctuations for finite N .

The effect of fluctuations: Large deviation theory

Key object: **Rate function**

$$S_T(\phi) = \frac{1}{2} \int_0^T \left| \sigma(\phi)^{-1} \left(\dot{\phi} - b(\phi) \right) \right|^2 dt = \frac{1}{2} \int_0^T \mathcal{L}(\phi, \dot{\phi}) dt$$

associated with the S(P)DE

$$dX^\epsilon(t) = b(X^\epsilon(t)) dt + \sqrt{\epsilon} \sigma(X^\epsilon(t)) dW(t)$$

Then, the probability that $\{X^\epsilon(t)\}_{t \in [0, T]}$ is close to a path $\{\phi(t)\}_{t \in [0, T]}$ is

$$\mathcal{P} \left\{ \sup_{0 \leq t \leq T} |X^\epsilon(t) - \phi(t)| < \delta \right\} \asymp \exp(-\epsilon^{-1} S_T(\phi))$$

for $\epsilon \rightarrow 0$. The problem is reduced to a **minimization** problem

$$\mathcal{P} \{X^\epsilon(T) \in A | X^\epsilon(0) = x\} \asymp \exp \left(-\epsilon^{-1} \inf_{\phi: \phi(0)=x, \phi(T) \in A} S_T(\phi) \right)$$

The effect of fluctuations: Large deviation theory

Propulsion

Gradient system

$$\partial_t \rho = -M \delta E / \delta \rho + \sqrt{\frac{2}{N}} M^{1/2} \eta(x, t)$$

for

$$E(\rho) = \int_{\Omega} (\rho \log \rho - \rho + f(\rho) + \frac{1}{2} \delta^2 |\nabla \rho|^2) dx$$

$$M(\rho) \xi = \nabla \cdot (\rho D(\rho) \nabla \xi)$$

The effect of fluctuations: Large deviation theory

Propulsion

Gradient system

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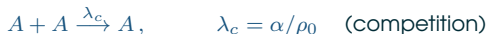
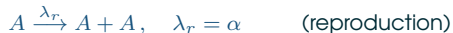
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Reproduction

Poisson processes at each location for the reactions



then LLN is

$$\partial_t \rho = \alpha \rho (1 - \rho / \rho_0)$$

with Poisson noise.

The effect of fluctuations: Large deviation theory

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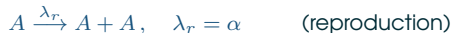
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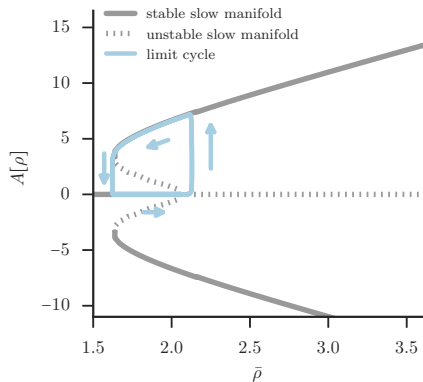
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(corresponding SPDE is ill-posed)

The effect of fluctuations: Quasi time periodic regime

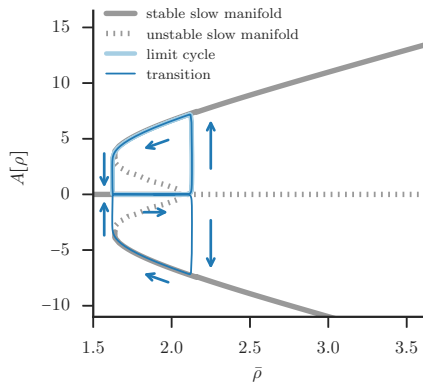


Consider $\rho_S < \rho_0 < \rho_c$

- Tiny fluctuations alter the structure of the limit cycle
- These are not rare events:
Zero action for transition when $\alpha \rightarrow 0$

T. Grafke, M. Cates, E. Vanden-Eijnden (2017) arXiv:1703.06923

The effect of fluctuations: Quasi time periodic regime

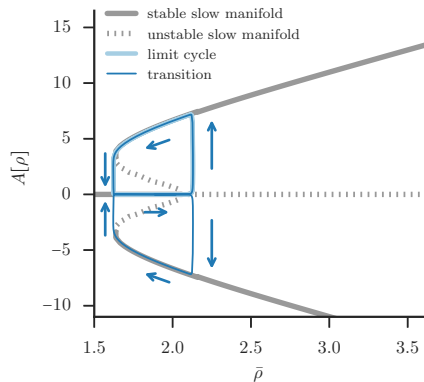


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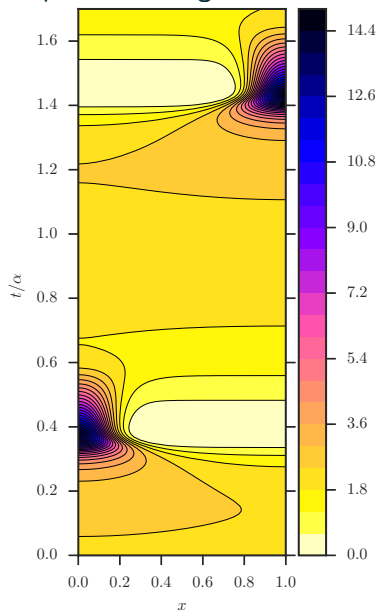
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The effect of fluctuations: Quasi time periodic regime



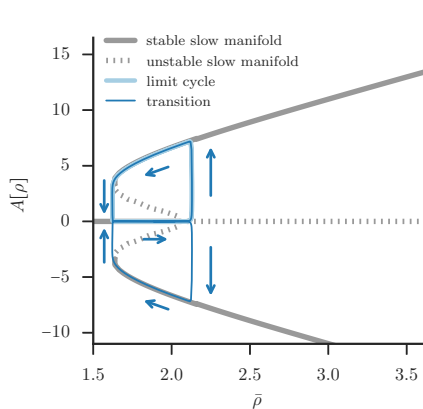
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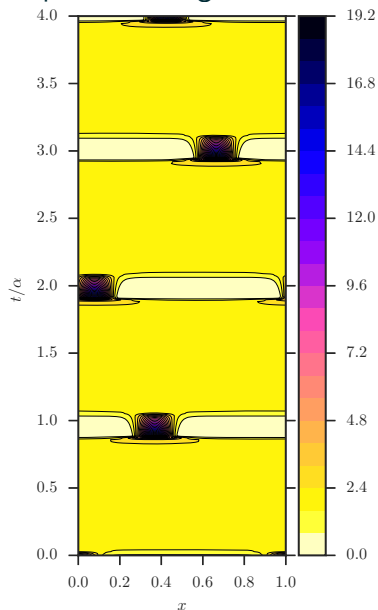
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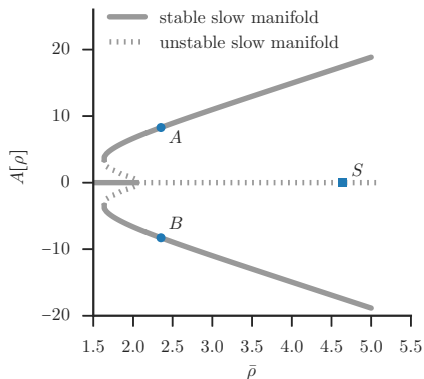
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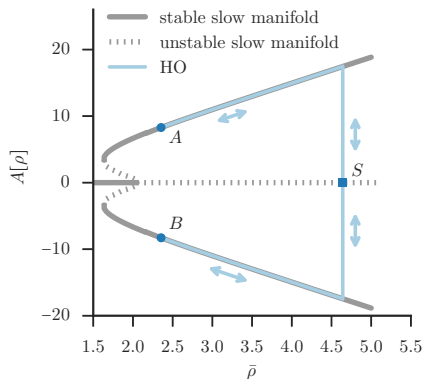
The effect of fluctuations: Metastable regime



Consider $\rho_c < \rho_0$

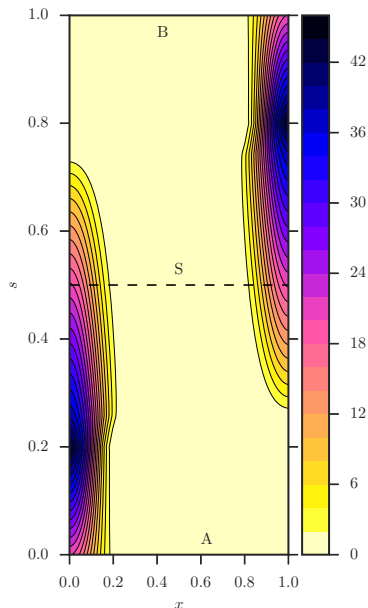
- Fixed points become **metastable**
- Transitions between them:
Finite action, exponentially small probability, *LDT regime*

The effect of fluctuations: Metastable regime



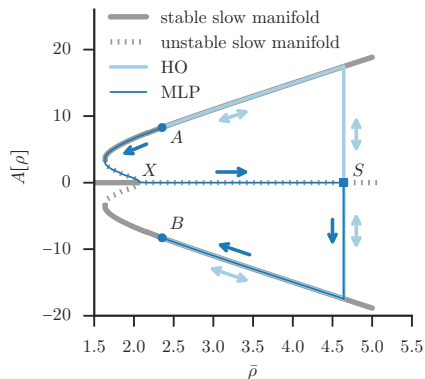
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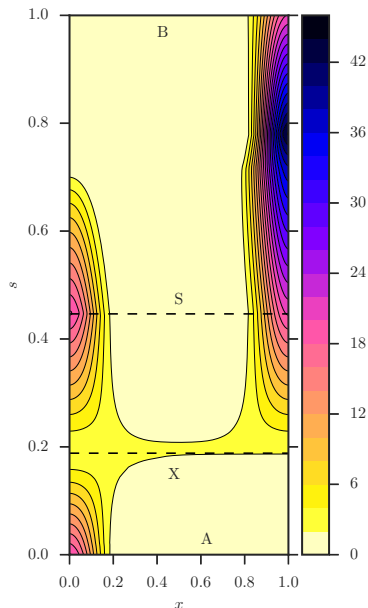
T. Grafke, E. Vanden-Eijnden (2017) arXiv:1704.06723

The effect of fluctuations: Metastable regime



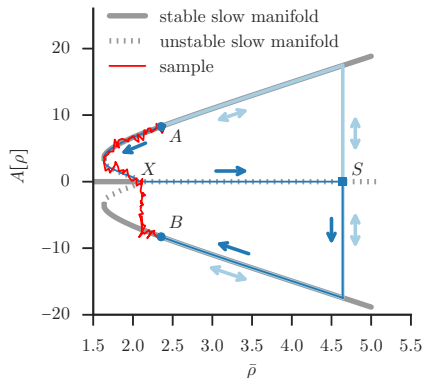
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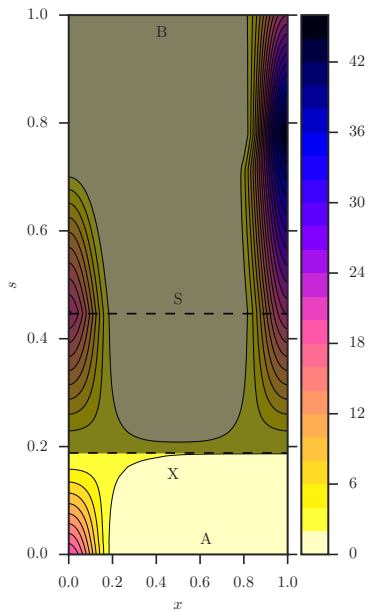


T. Grafke, E. Vanden-Eijnden (2017) arXiv:1704.06723

The effect of fluctuations: Metastable regime



At finite N , accounting for entropy, only part of the transition with **non-zero action** is robust (and matters).

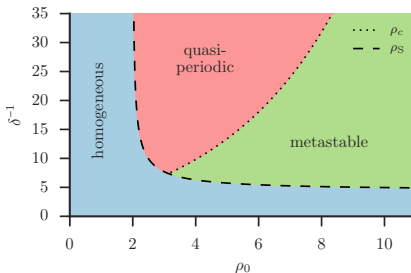
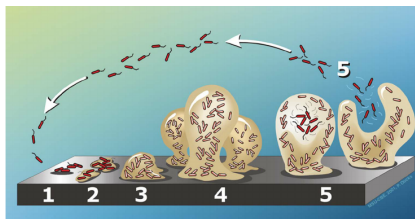


Non-equilibrium self-organization

Complex collective behavior for simple active agents:

Propulsion and Reproduction

- When $\rho_0 < \rho_S$, **planktonic** phase is robust.
- When $\rho_S < \rho_0 < \rho_c$, particles oscillate between **biofilm** and **planktonic** phase
- When $\rho_0 < \rho_c$, biofilms are **metastable**. They **rarely** disperse and reform by dying out
- Full **phase diagram** depends on carrying capacity ρ_0 and **domain size** δ^{-1} .



T. Grafke, M. Cates, E. Vanden-Eijnden (2017) arXiv:1703.06923

Numerical Computation of Large Deviation Minimizers

Main problem

For our system and observable, find the **minimizer** ϕ^* such that

$$S_T(\phi^*) = \inf_{\phi} S_T(\phi),$$

where the minimization is over all trajectories fulfilling the boundary conditions.

The knowledge of this **minimizer** (MLP) yields

- Most probable **evolution** in time from initial state into this final configuration
- Corresponding **optimal force**, computable from auxiliary field θ
- Tail scaling behavior of the **PDF** of our observable, roughly through $\mathcal{P}\{x^\epsilon(T)\} \sim \exp(-\frac{1}{\epsilon} \inf_{\phi} S_T(\phi))$

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Its **computation** is challenging:

- This is an infinite-dimensional **PDE constraint optimization**. The search space is large (space-time).
Each **iteration** we have to solve a system of coupled PDEs.
- If we are computing **transition probabilities**, we are interested in the **Quasipotential**,

$$V(x_1, x_2) = \inf_{T>0} \inf_{\phi} S_T(\phi)$$

This infimum is not attained in general, $T \rightarrow \infty$.

Challenges: Infinite transition time and geometric rate function

Quasipotential

$$V(x_1, x_2) = \inf_{T>0} \inf_{\phi} S_T(\phi)$$

This infimum is not attained in general, $T \rightarrow \infty$.

In the case $T \rightarrow \infty$, realize, that $\mathcal{H}(x, \theta) = 0$, so that

$$\int \mathcal{L}(x, \dot{x}) dt = \int \sup_{\theta} (\langle \dot{x}, \theta \rangle - \mathcal{H}(x, \theta)) dt = \sup_{\theta: \mathcal{H}(x, \theta) = 0} \int \langle \dot{x}, \theta \rangle dt$$

Effectively:

Reduce minimization over all paths to finding **geodesic** of the associated (Finsler) **metric**.

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Summary

- Non-equilibrium statistical mechanics theory of active matter is still in its infancy
- One example is **Motility induced phase separation** — direct consequence of motile agents with density dependent drift velocity
- Adding **population dynamics** is enough to yield complex emergent behavior reminiscent of biofilm-planktonic lifecycle
- Fluctuations in MIPS + Reproduction can be analyzed by LDT
- Noise-driven spatio-temporal self-organization
 - **Limit cycles** are temporally but not spatially robust against fluctuations
 - Transitions between **metastable colonies** are out-of-equilibrium and occurs different to detailed-balance intuition
- Computation via **geometric minimization** of LDT rate function