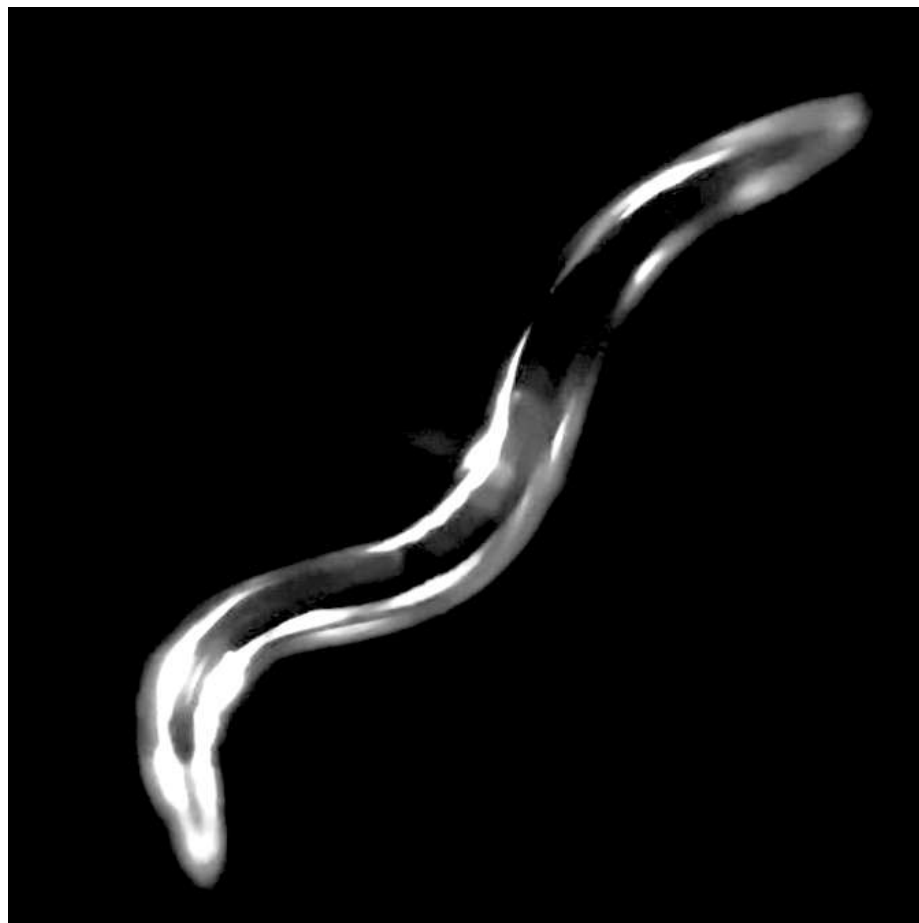
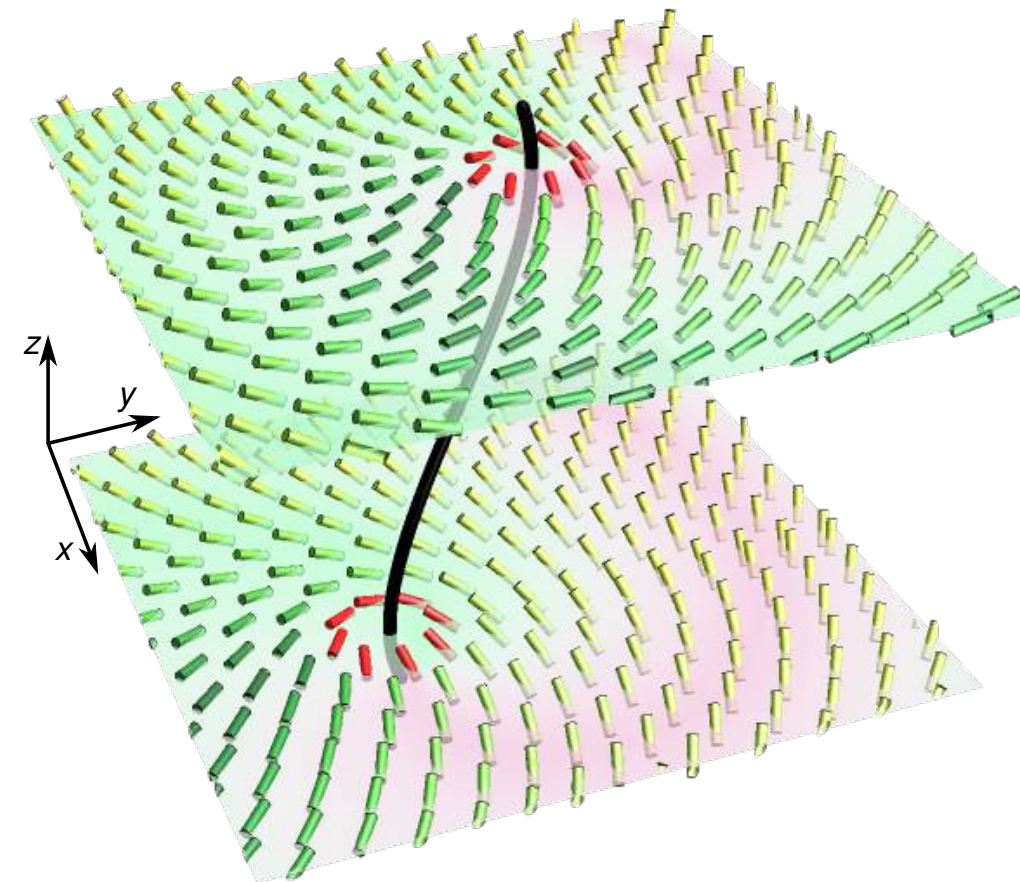


Quantum analogues in living and soft matter

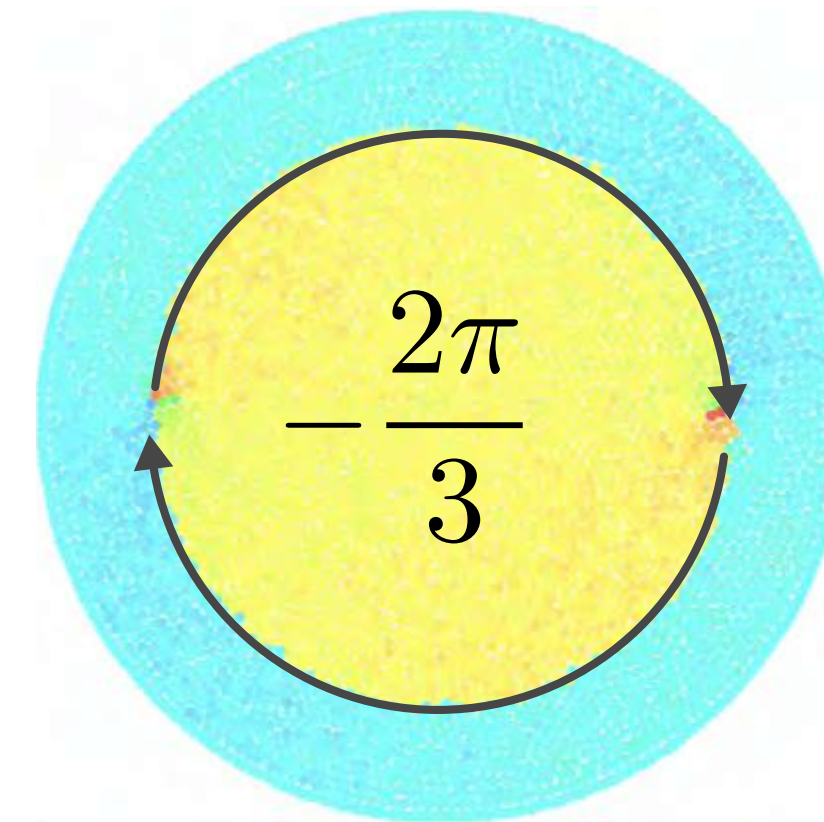


Flavell lab (MIT BCS)

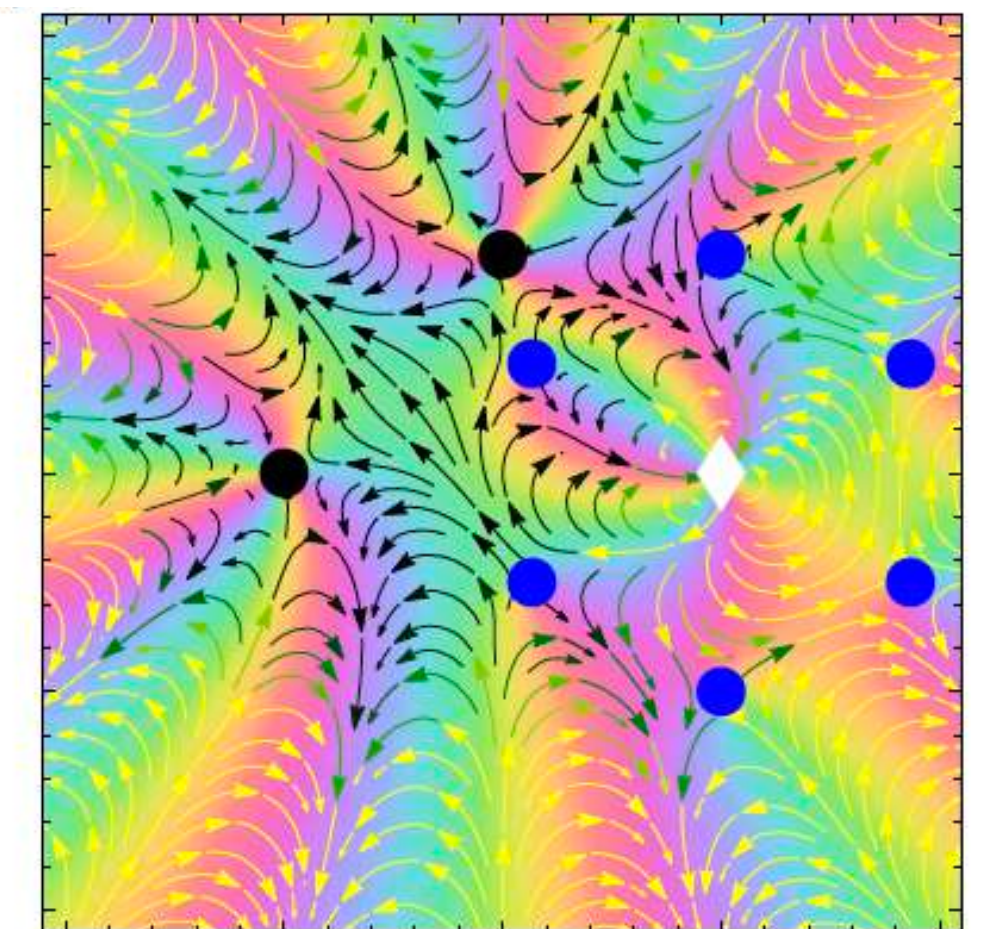
PRL 2023



Science Advances 2022



PRX 2022



$$\frac{1}{\sqrt{2}}|11\rangle \otimes (|0\rangle + |1\rangle)$$

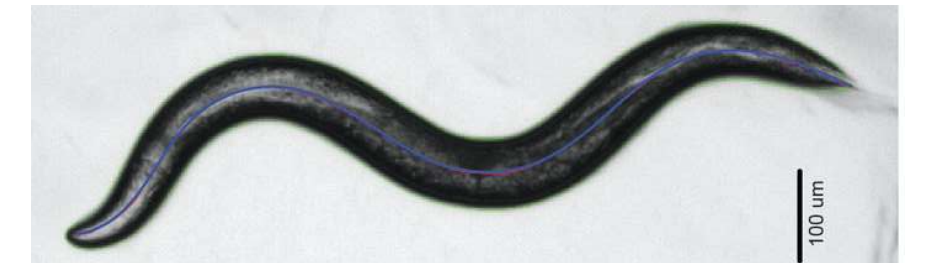
PRR 2024



Outline

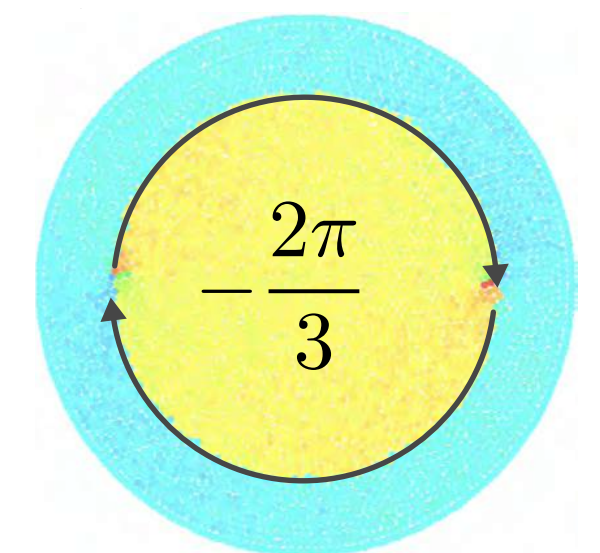
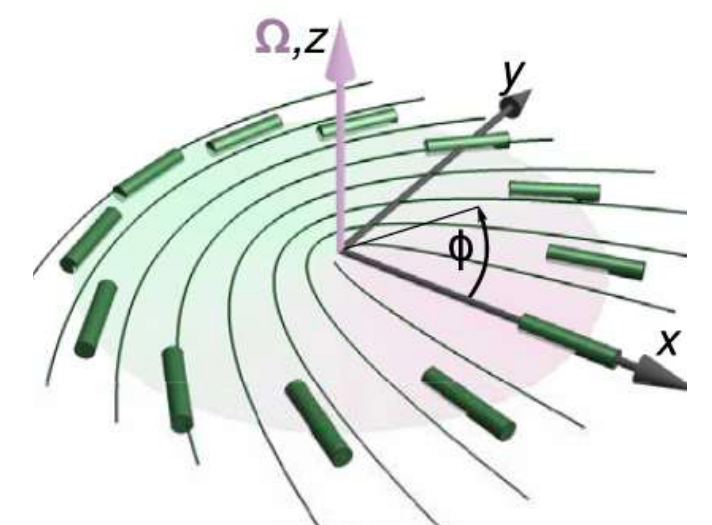
1. Dynamics of quasi-1D living systems

- spectral representation & analysis
- effective Schrödinger-like short-time dynamics



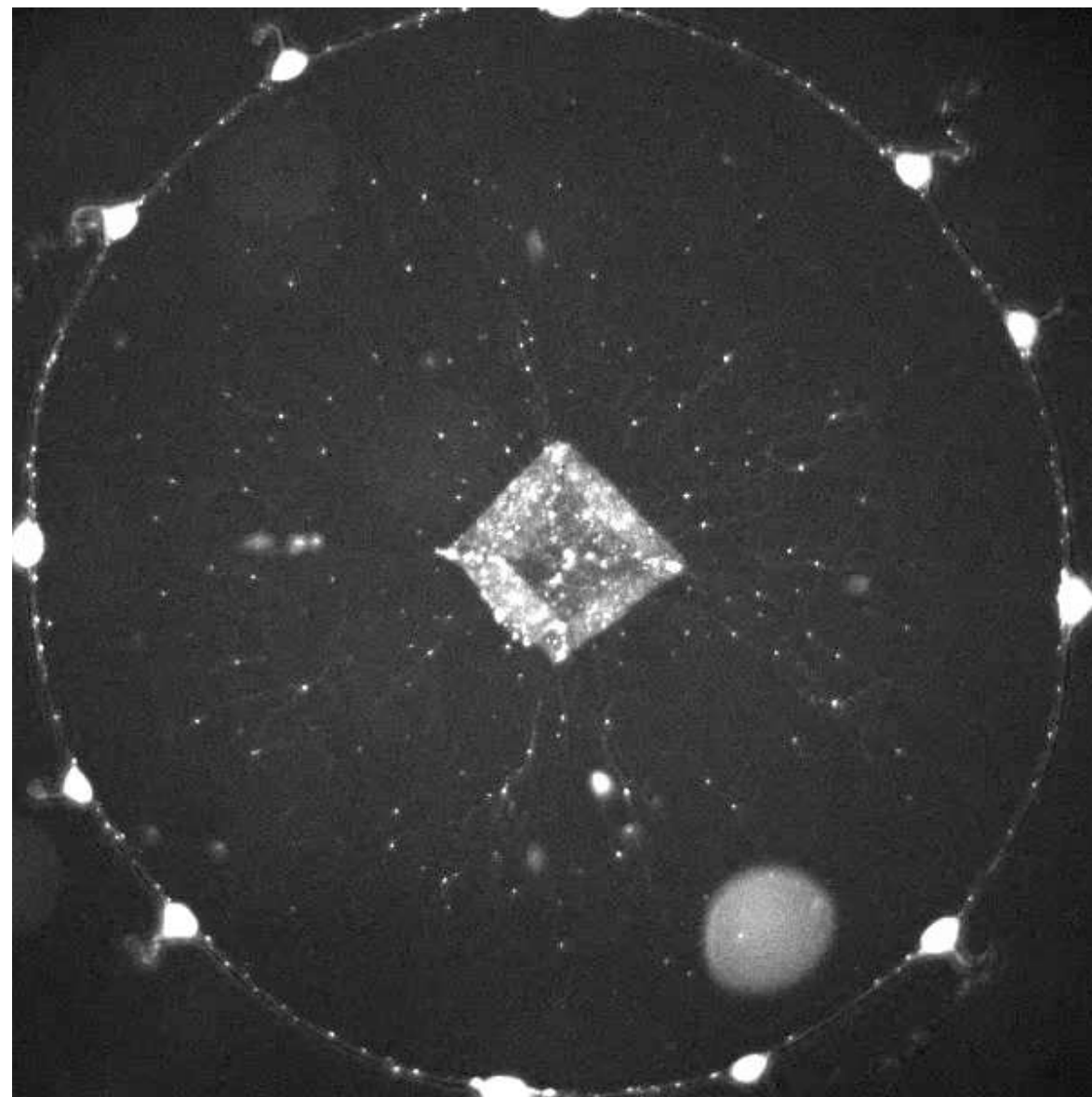
2. Topological defects & computation in liquid crystals

- nematic bits
- fractional defects and ‘anyonic’ braiding



How do living and soft matter systems compute ?

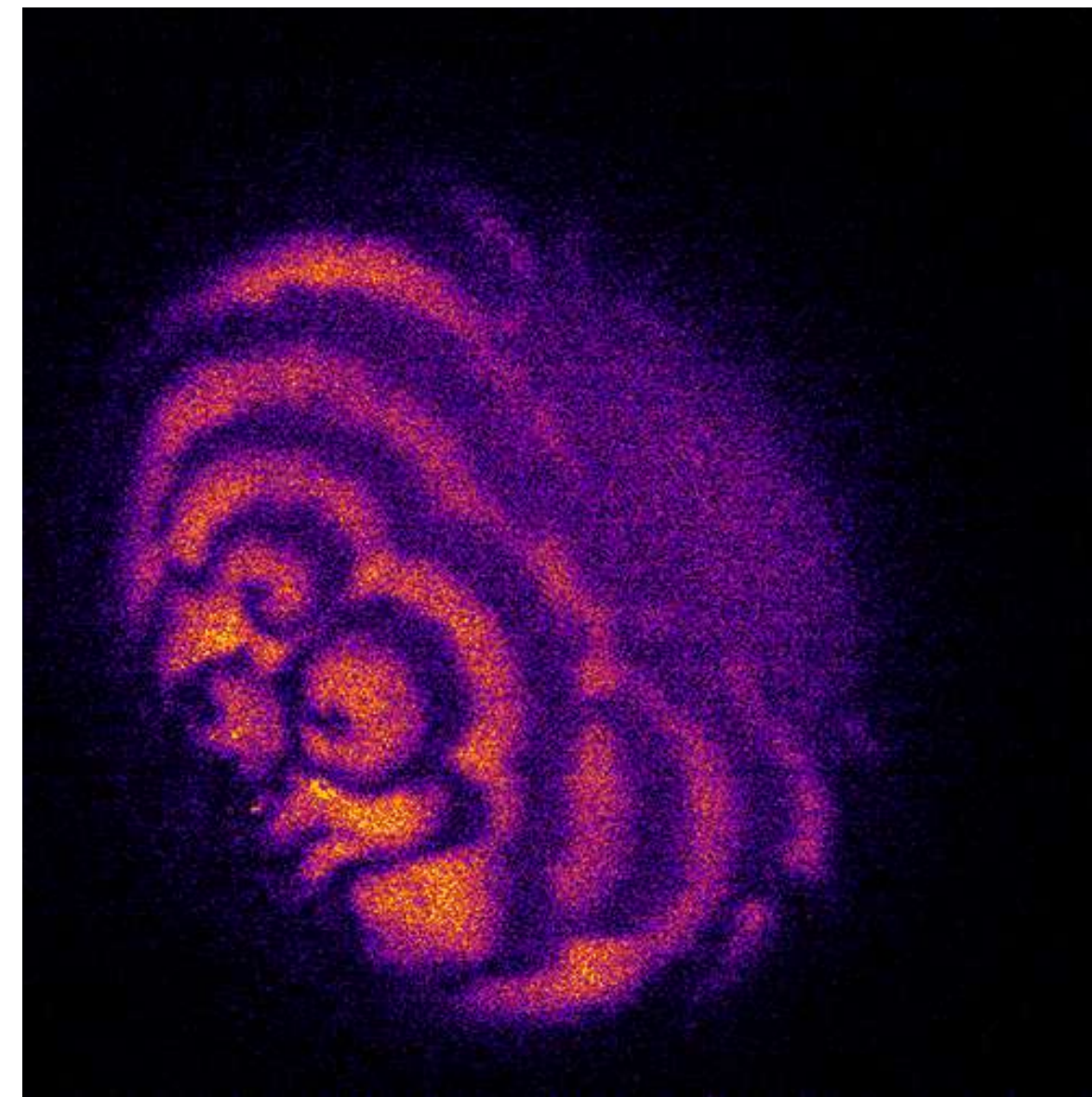
Jellyfish



Weissbourd lab (MIT Biology)

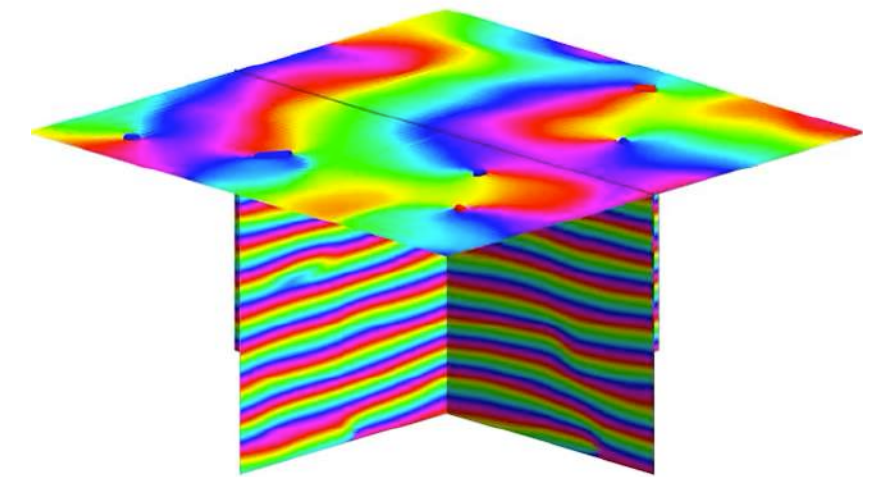
In progress

Starfish oocyte



Fakhri lab (MIT Physics)

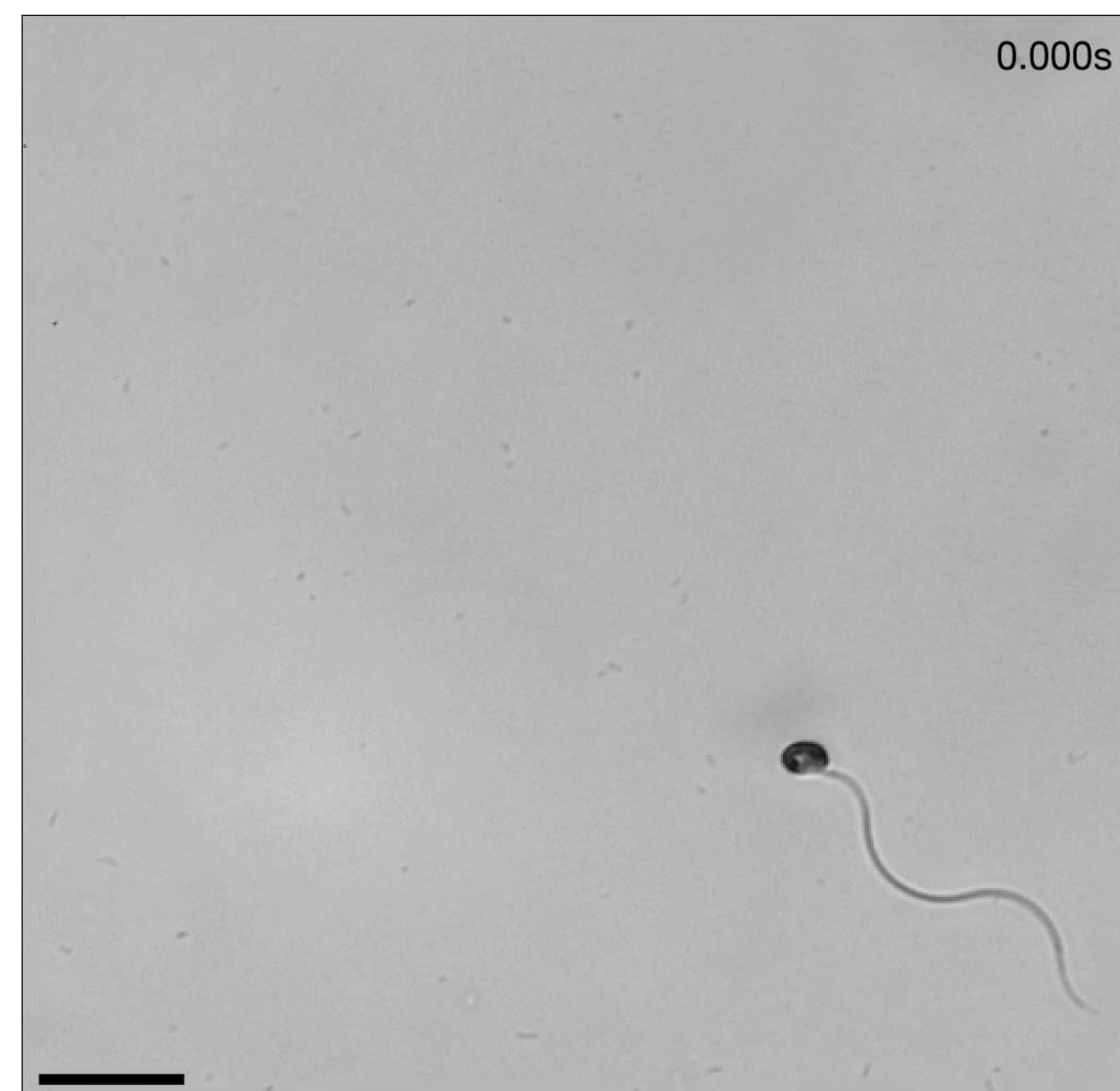
Nature Physics 2020
PNAS 2021



Part 1:

Spectral dynamics of quasi-1D living matter in 2D

Pterosperma alga



Wang Lab (Exeter Uni)

μm

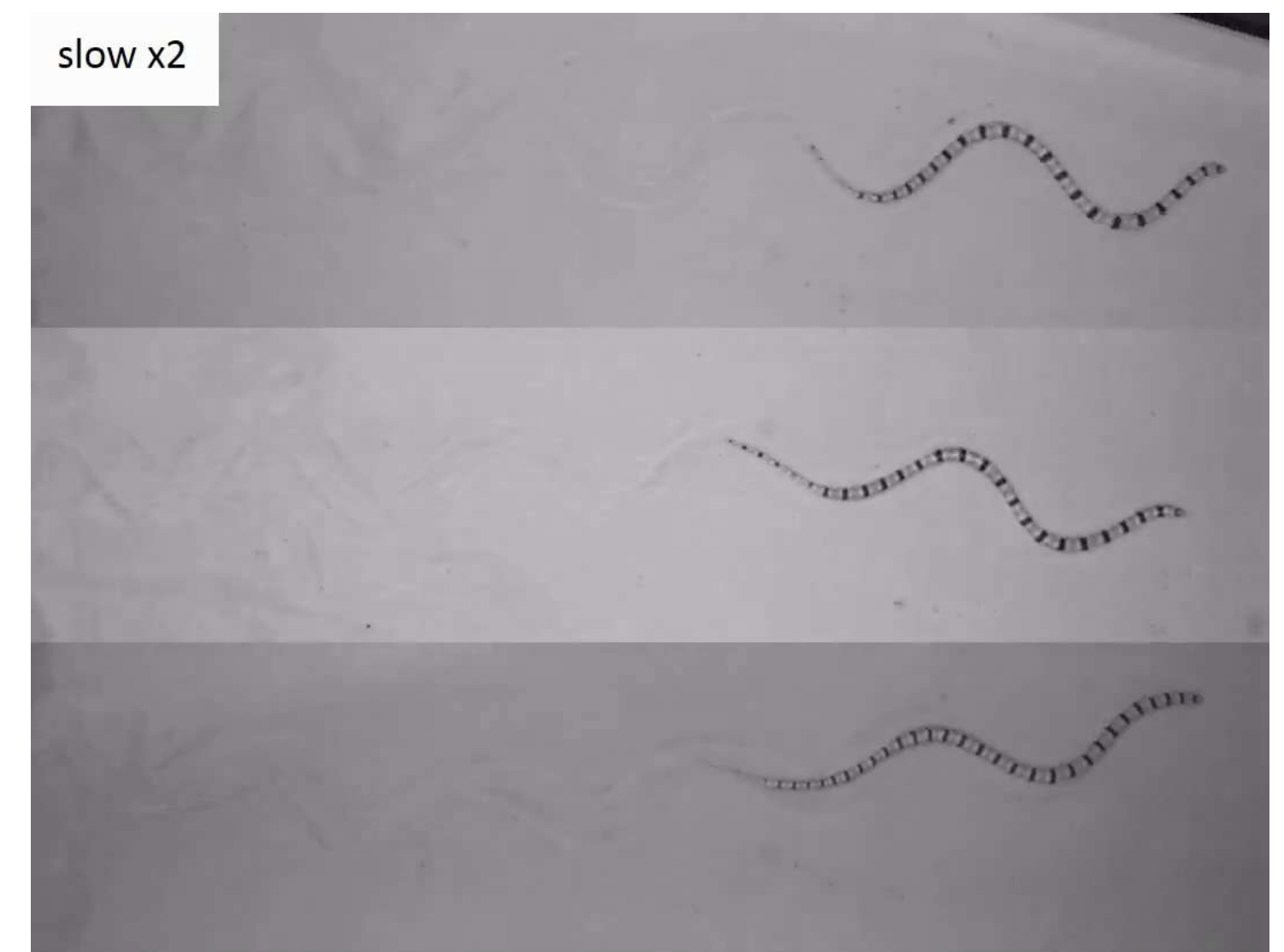
C elegans worm



Flavell Lab (MIT BCS)

mm

Desert snake



Goldman Lab (Georgia Tech)

m

Would like to ...

- Identify suitable collective ('slow') variables (low-dimensional representations)
- Identify and characterize different biological / biophysical states
- Infer dynamical models directly from video data
- Link biological signaling to physical dynamics

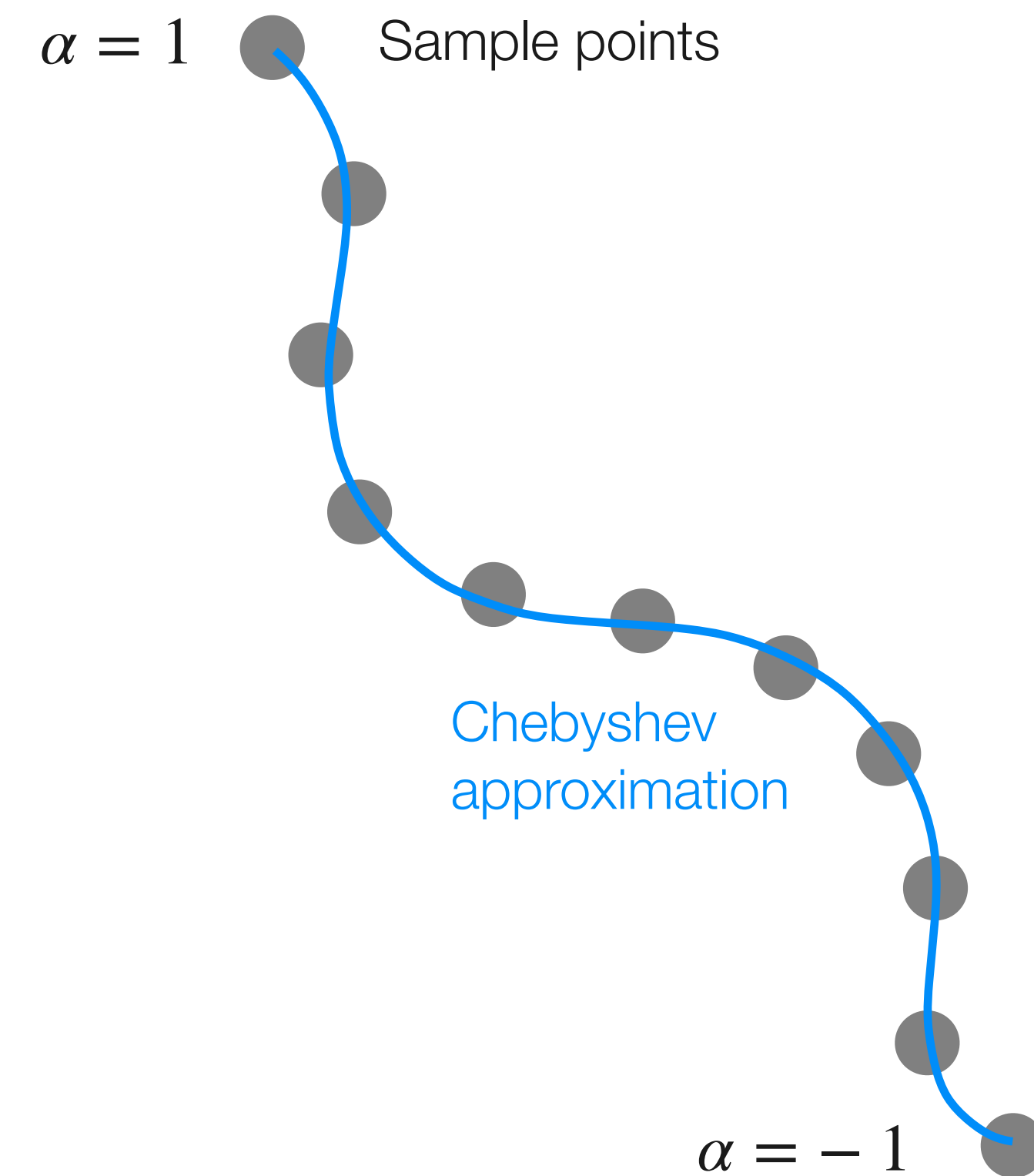
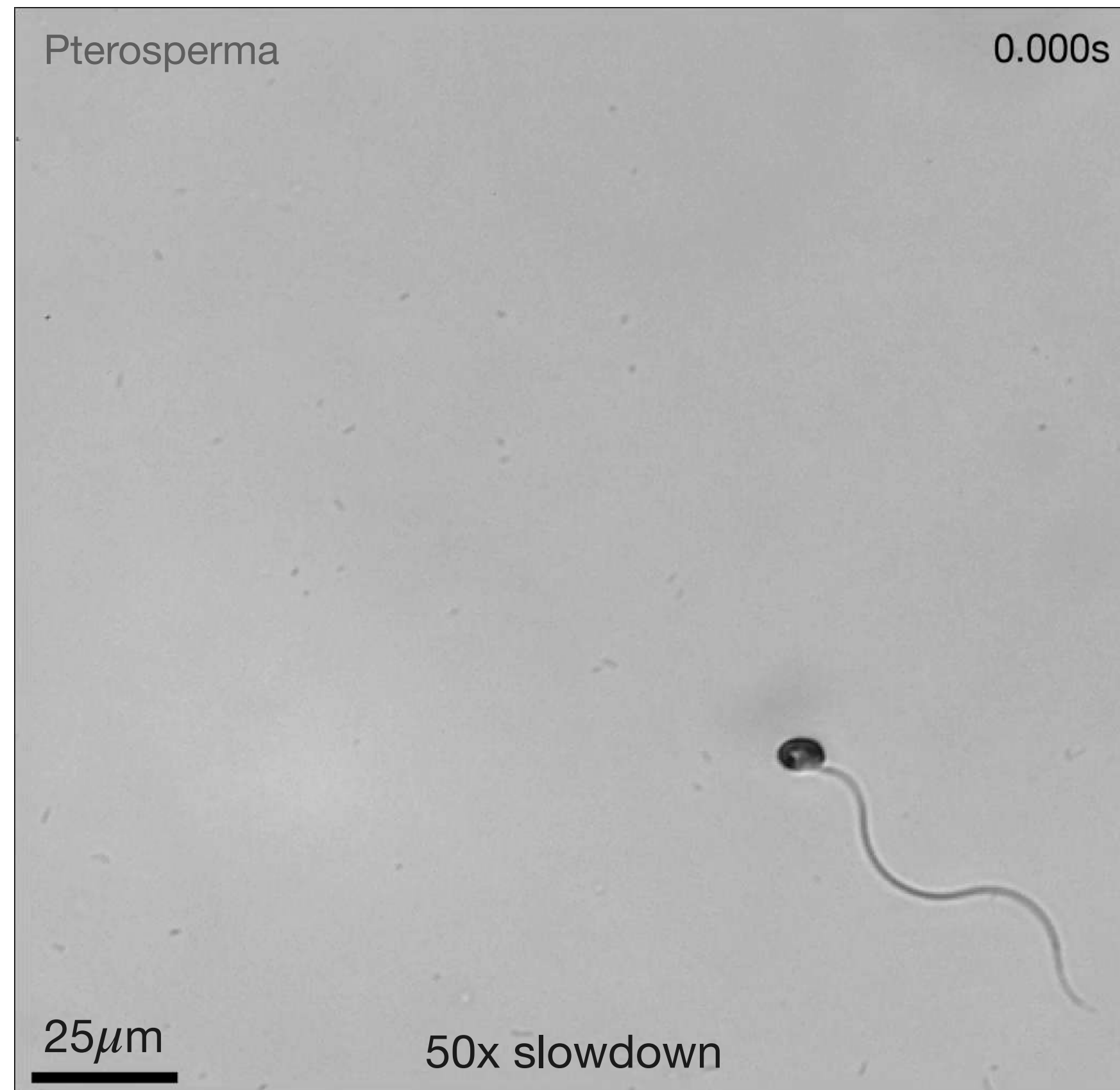
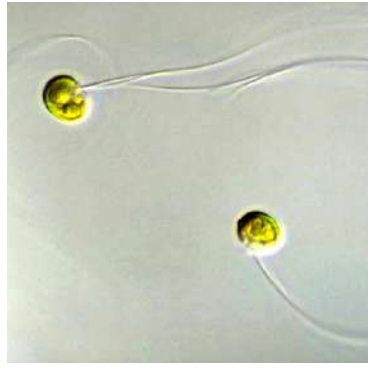


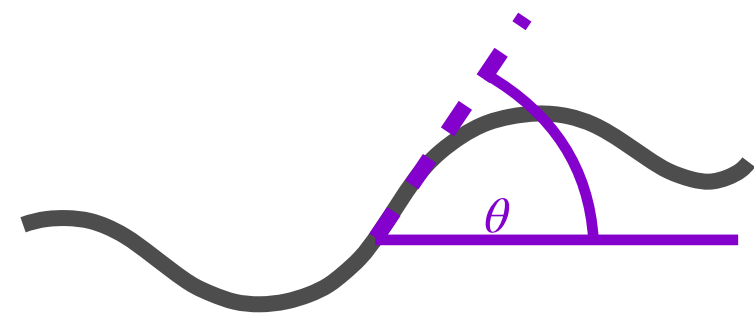
Kirsty Wan



Alex Boggon

Spectral representation of biological dynamics & behaviors

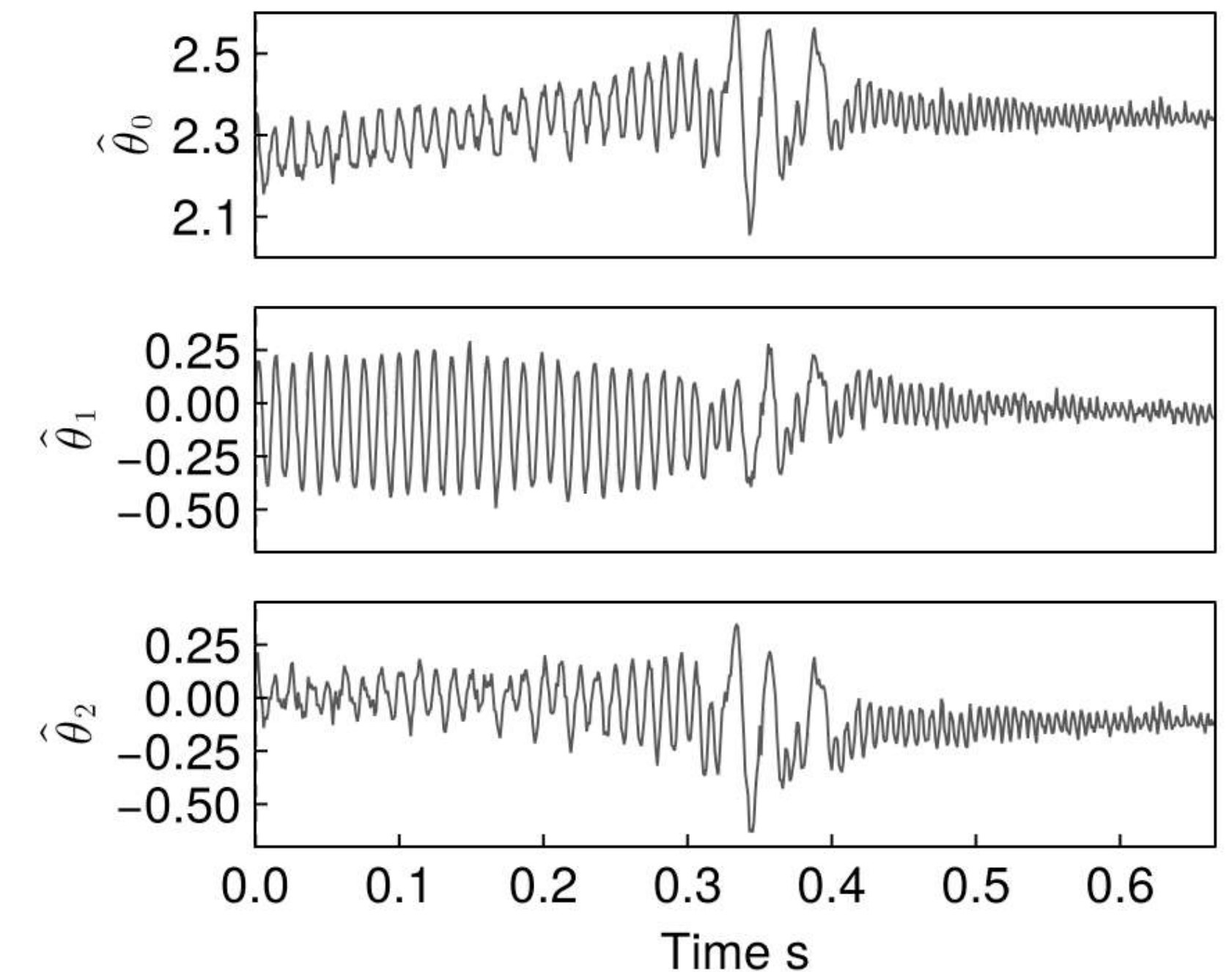
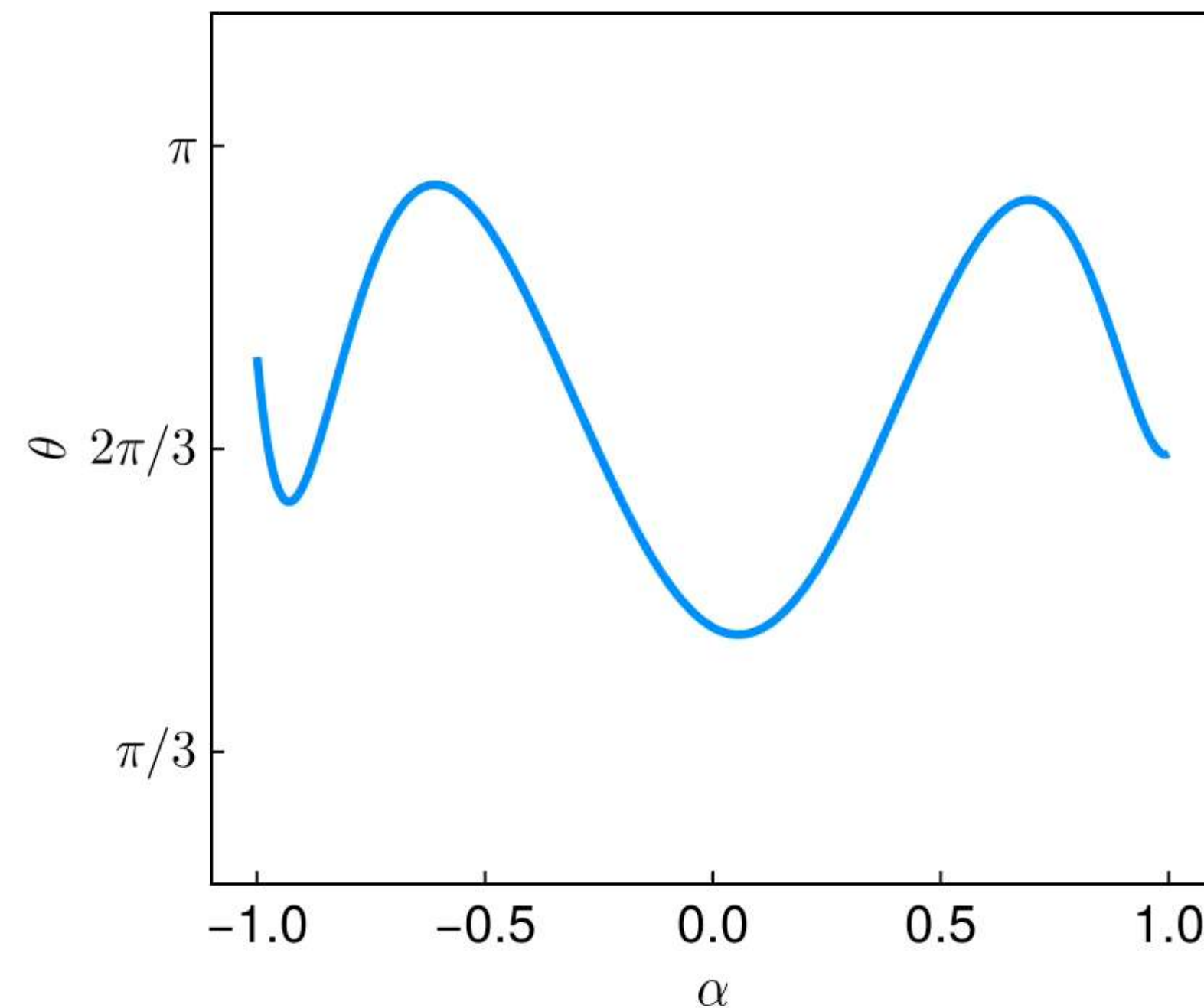
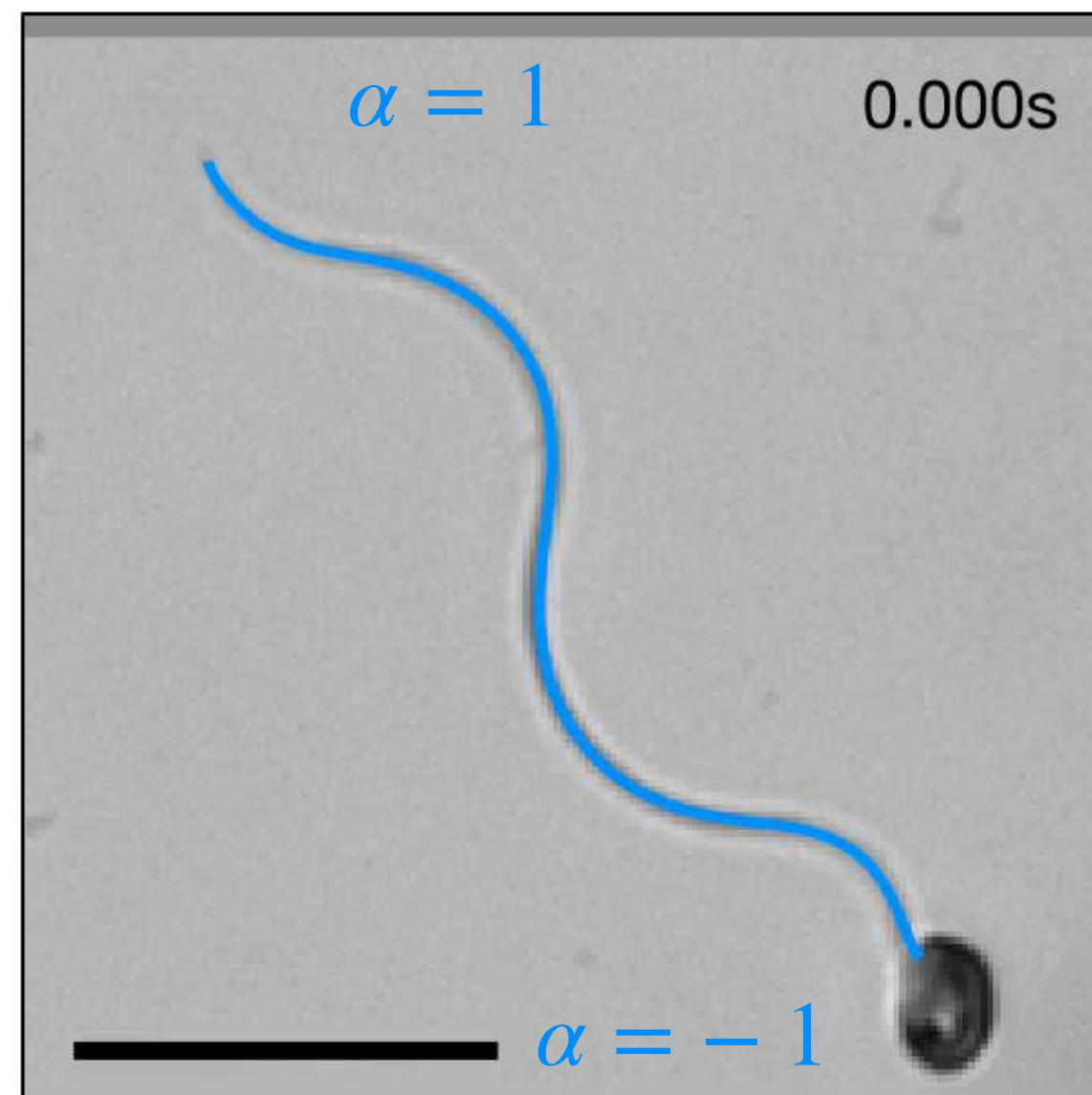




Spectral representation of biological dynamics & behaviors

$$\theta(t, \alpha) \approx \sum_{n=1}^N \hat{\theta}_n(t) T_n(\alpha) = \hat{\theta}_1(t) \text{ (blue line) } + \hat{\theta}_2(t) \text{ (orange U) } + \hat{\theta}_3(t) \text{ (green wavy) } + \hat{\theta}_4(t) \text{ (pink W) } + \dots$$

Orthogonal polynomial basis



Sparse 10-mode representation

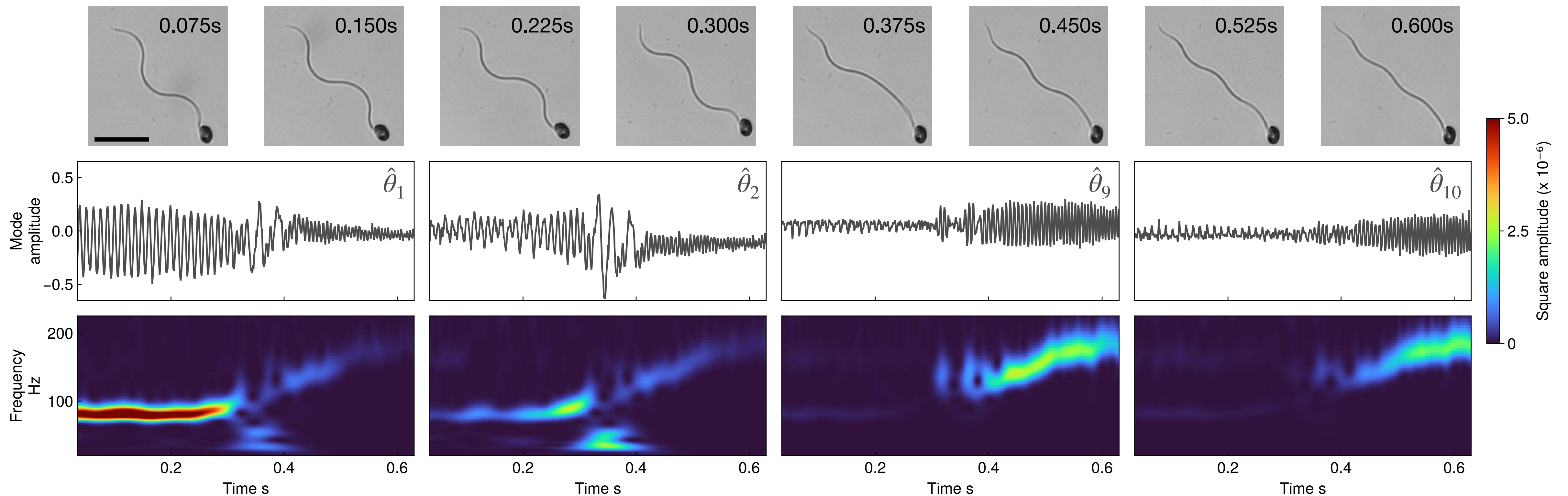
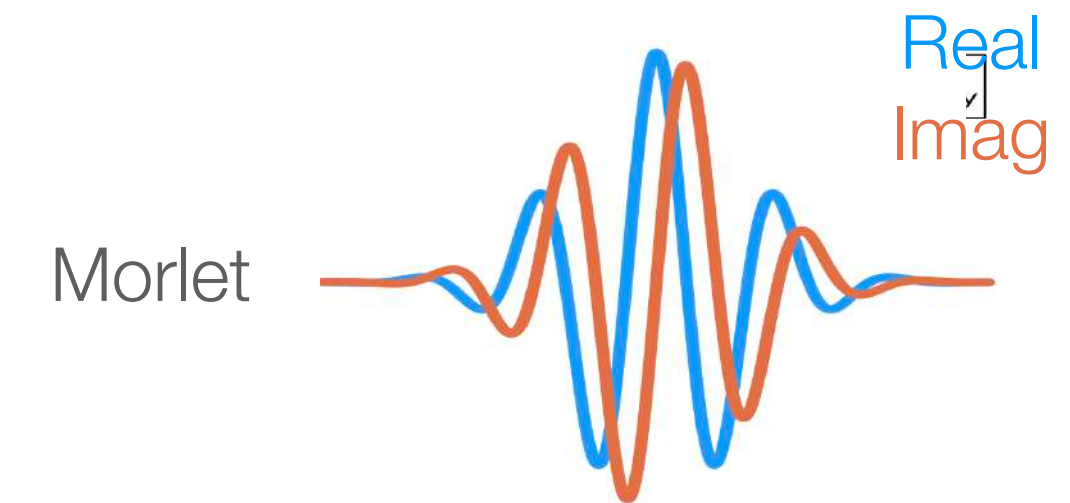
Wavelet analysis in time

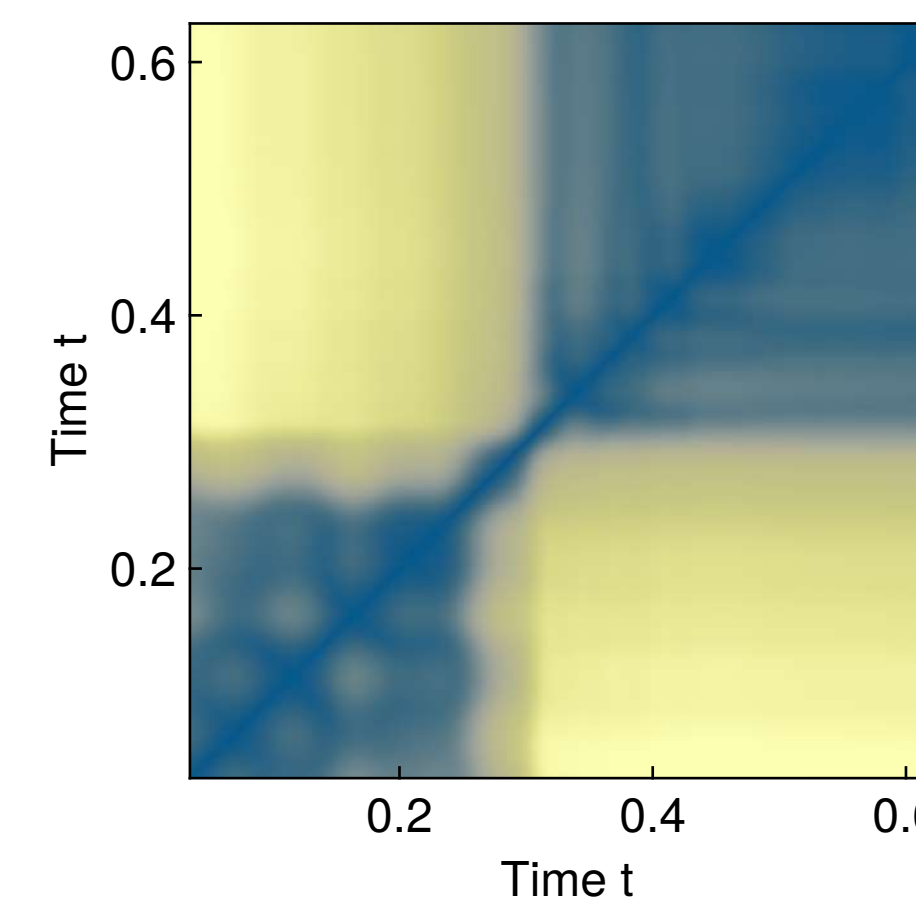
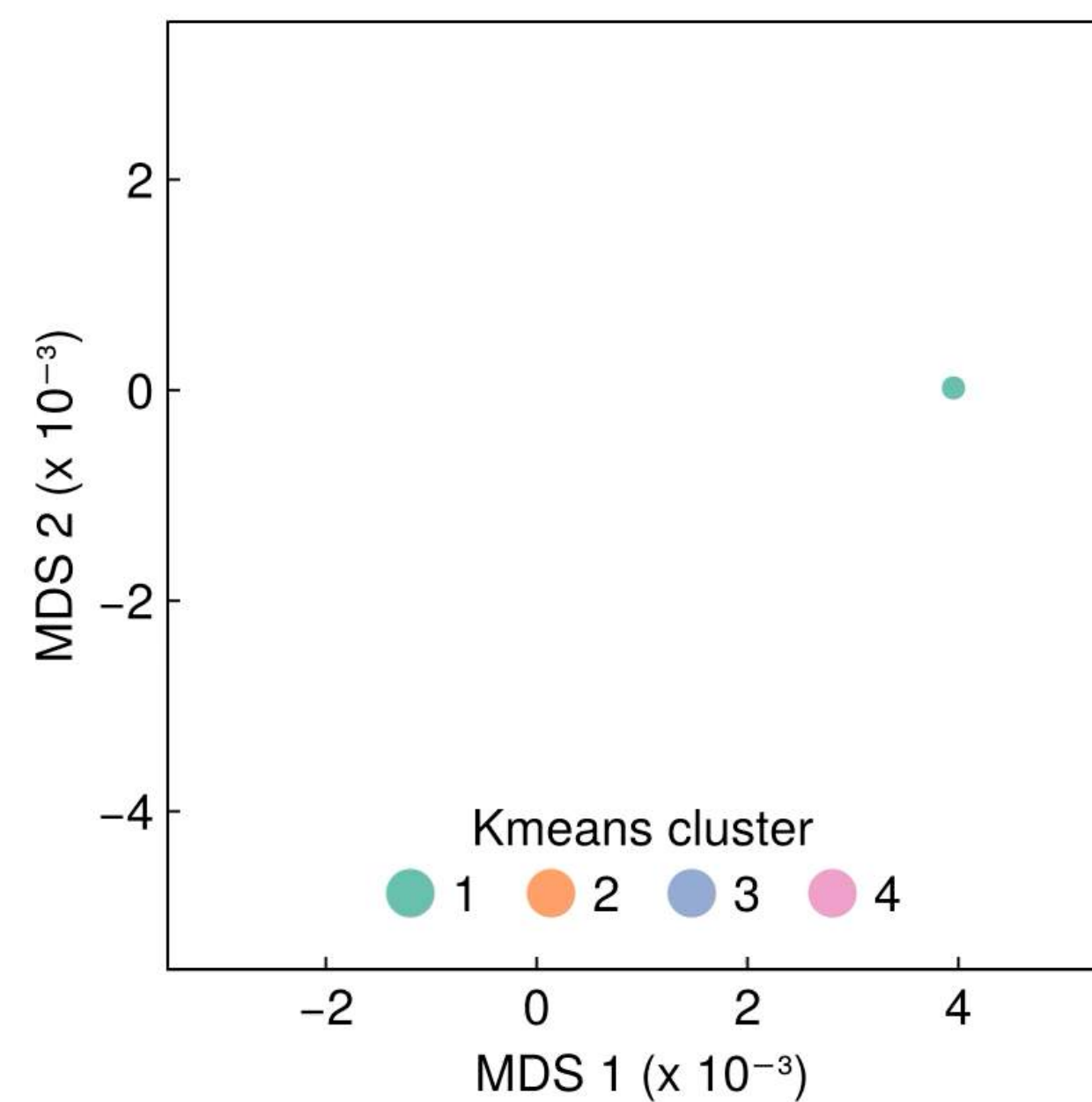
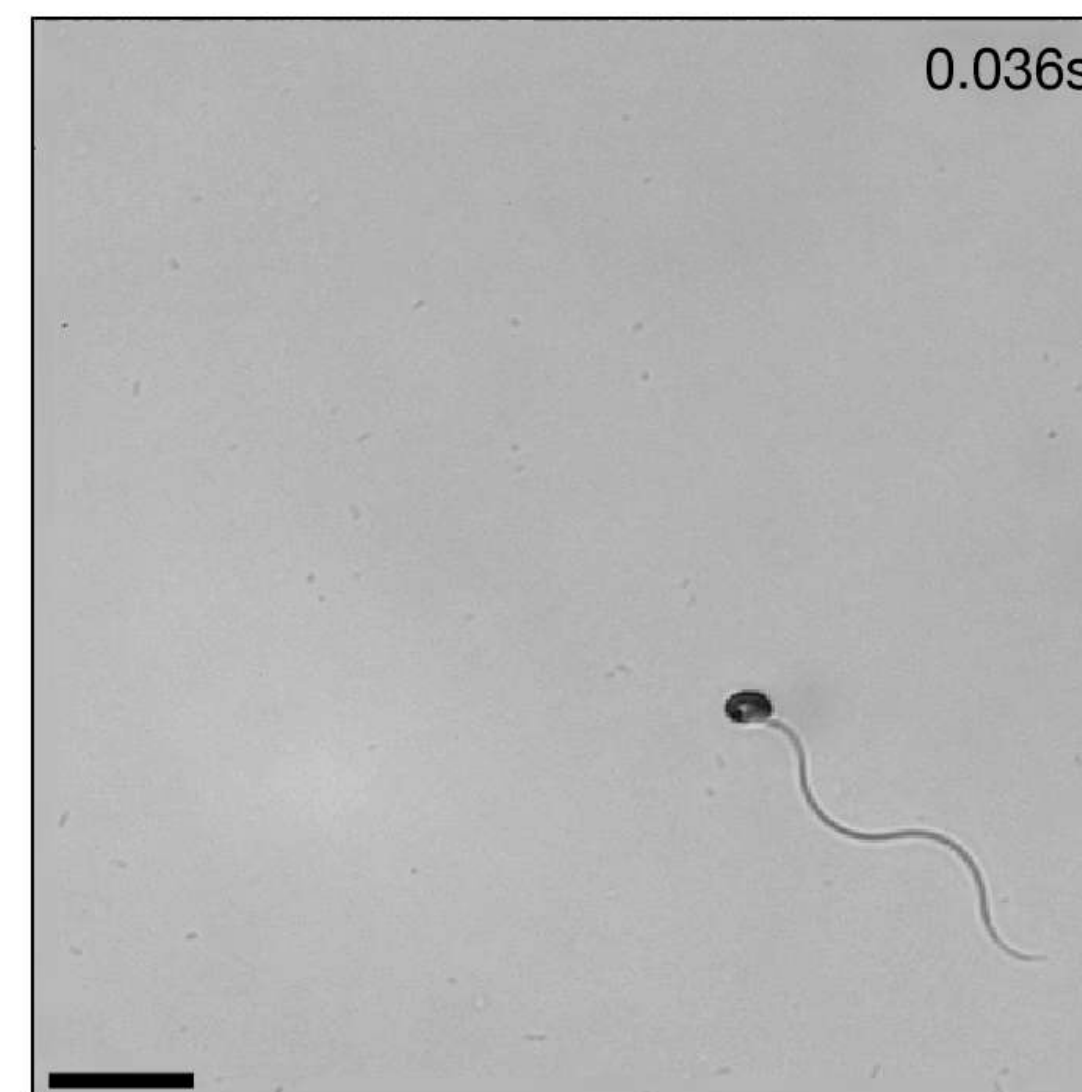
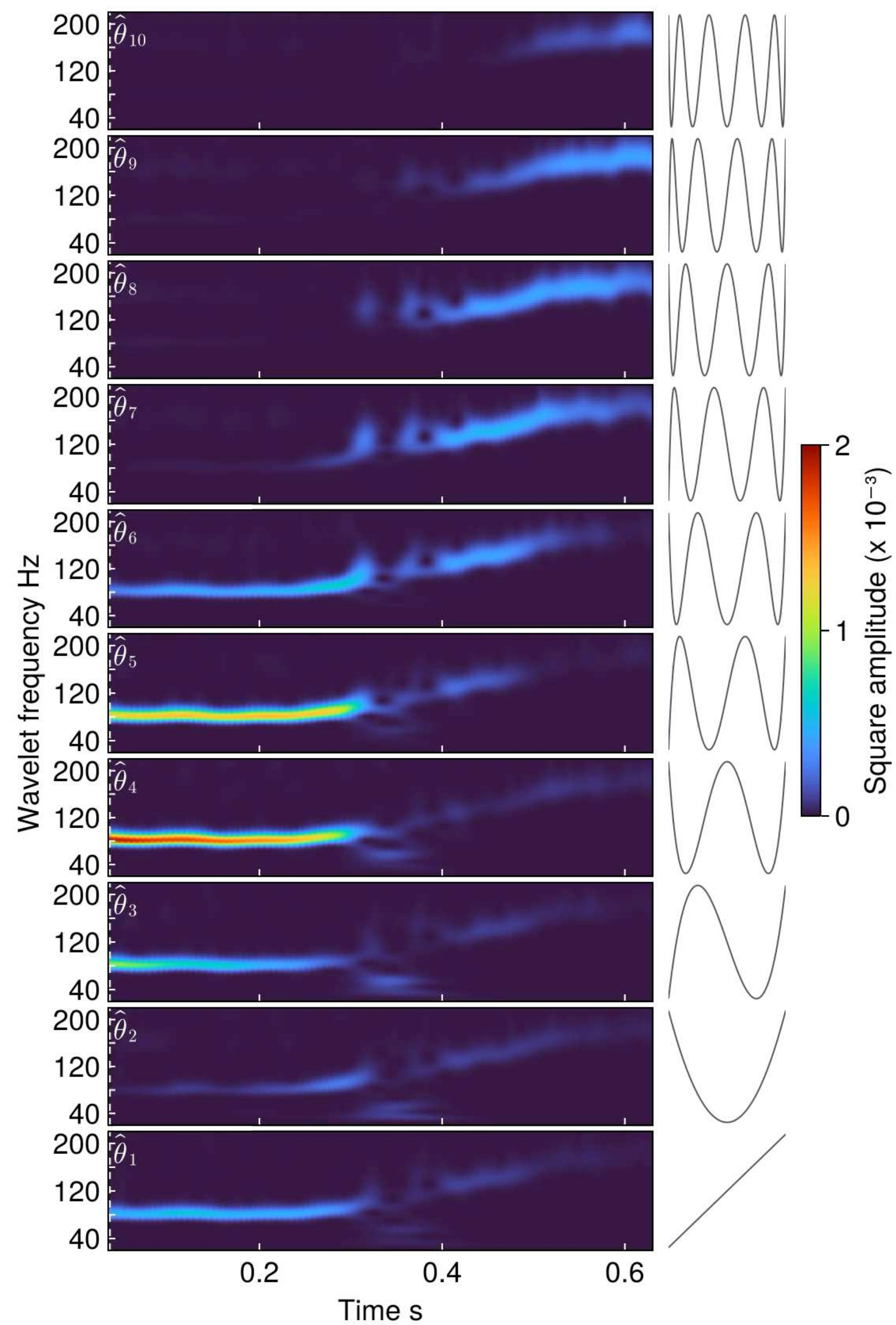
Convolve signal with shifts and scales of wavelet

$$W_{\psi}[f](a, b) = \frac{1}{|a|} \int dt f(t) \bar{\psi} \left(\frac{t - b}{a} \right)$$

wavelet frequency

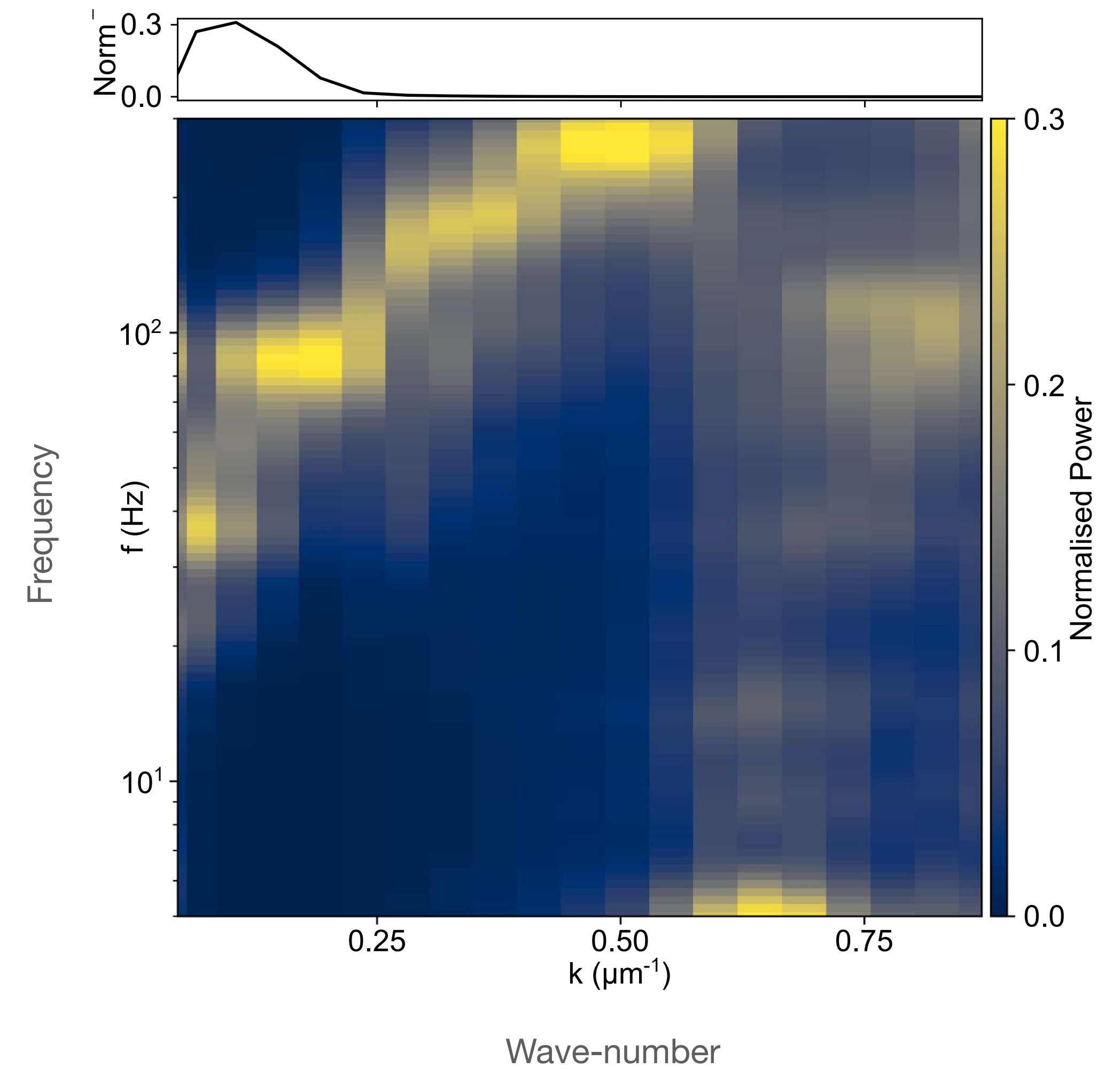
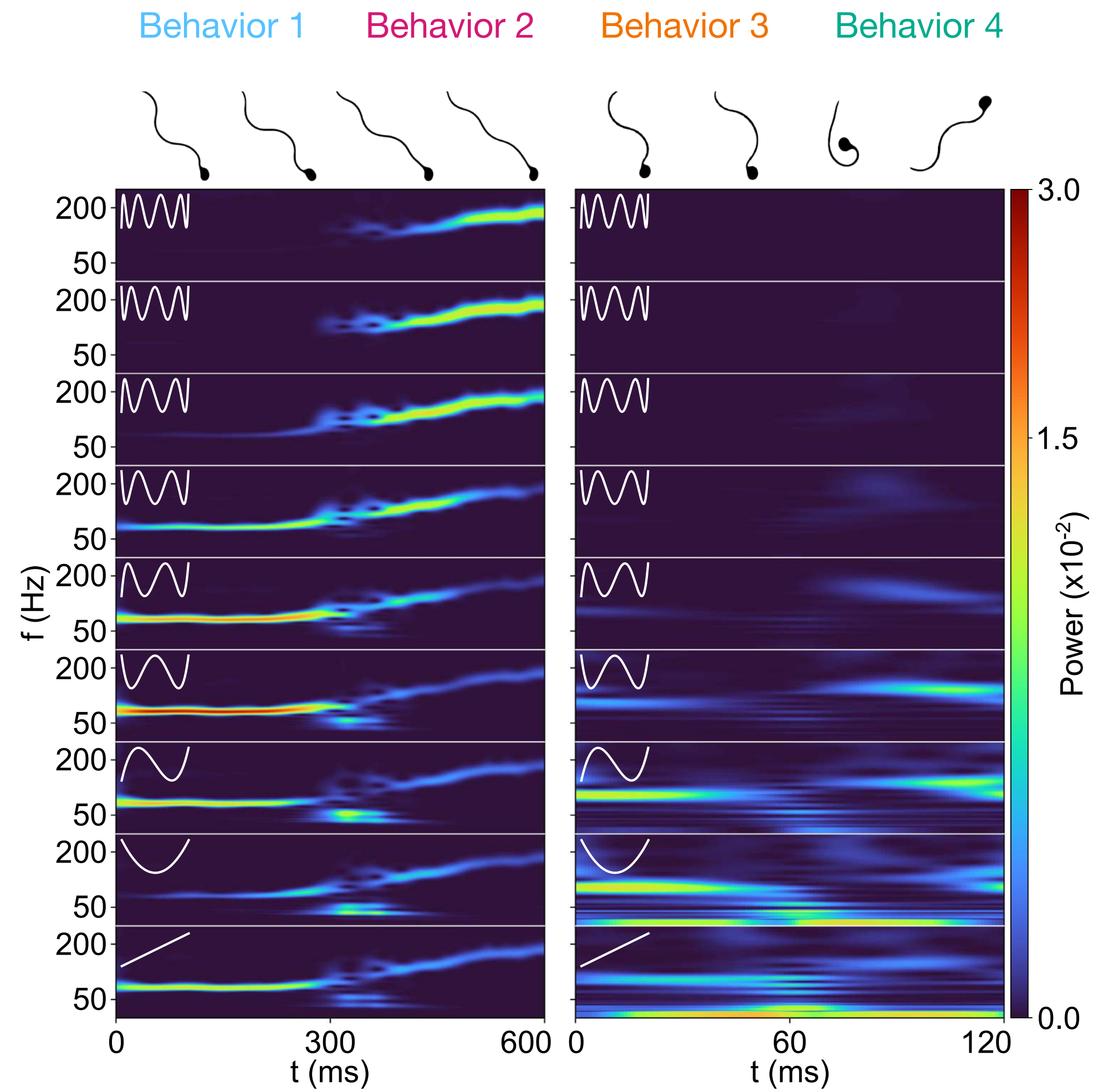
temporal
location





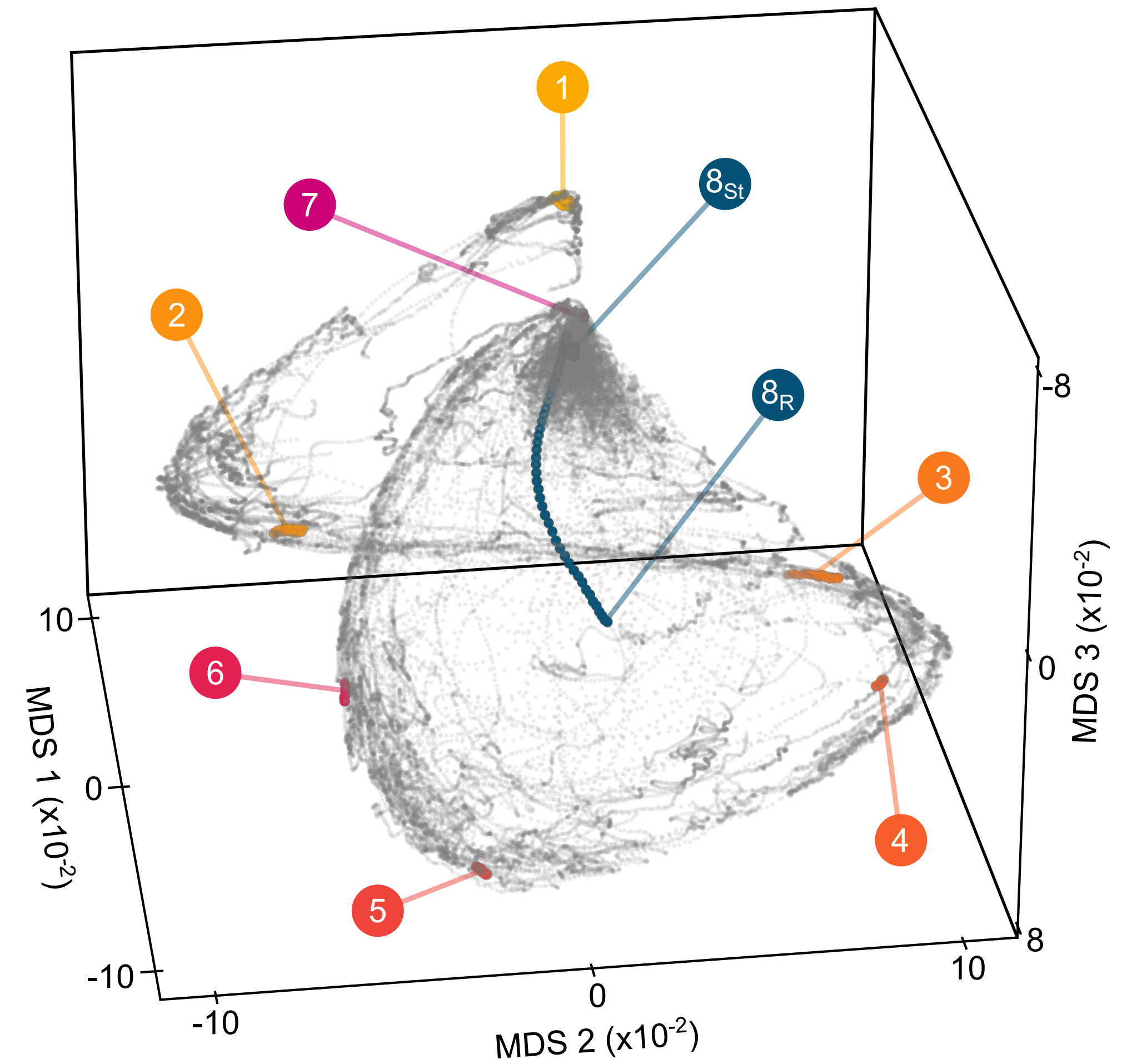
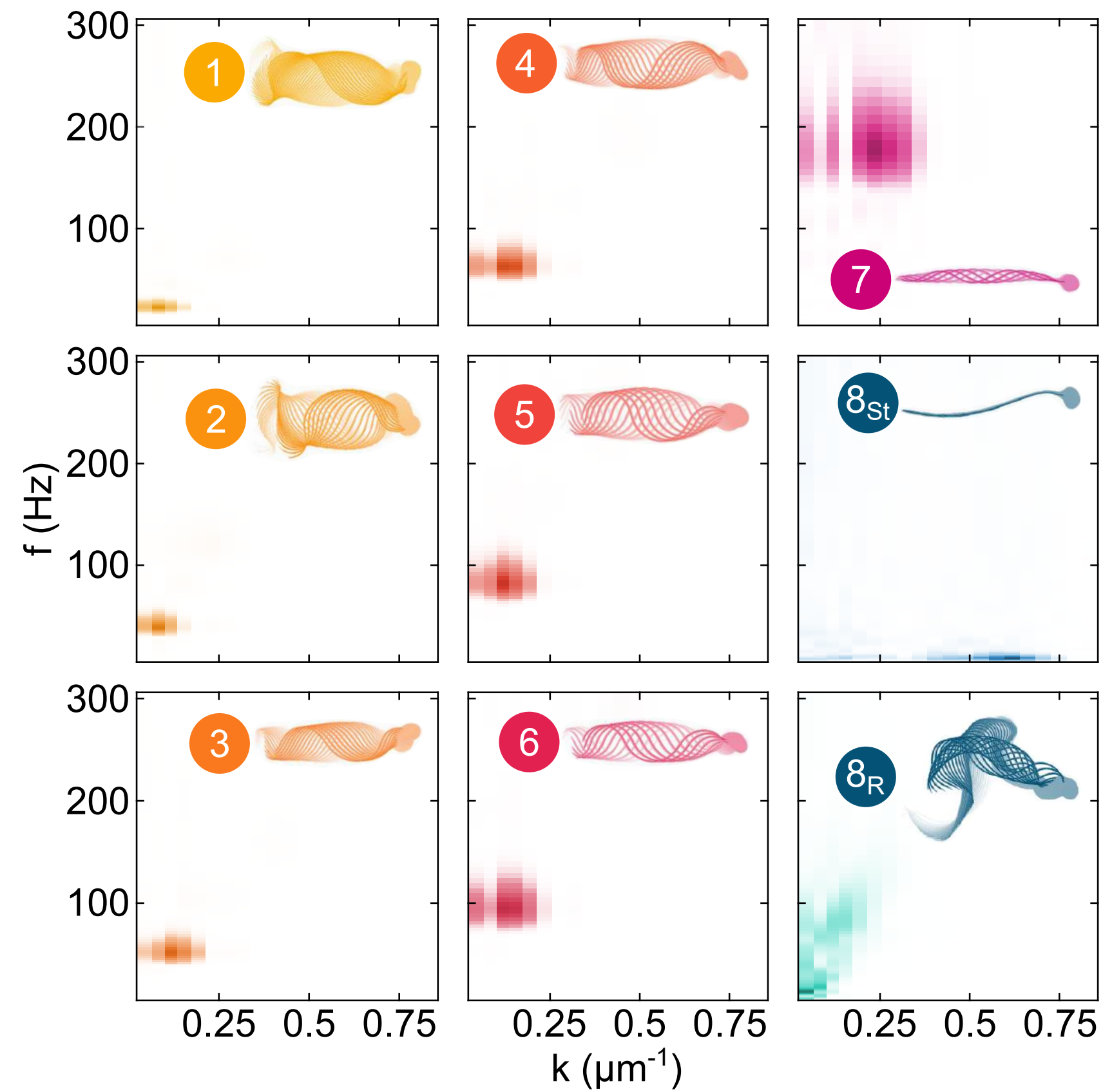
Instantaneous wavelet state
encodes
instantaneous dynamical behavior

Dispersion relation



Behavioral manifold

Wave-modes
activated
in a specific
behavioral state



Low-dimensional
embedding
(MDS based on
distances
between stacked
wavelet-vectors)

Generalizes across systems & scales

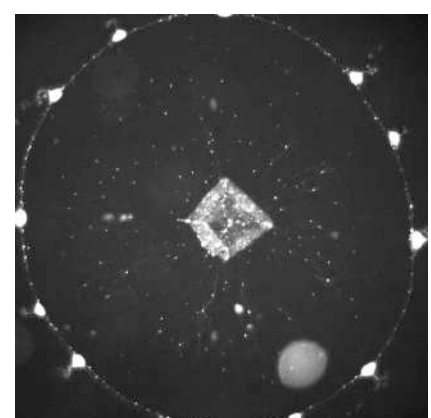
Roundworm



Legendre / Chebyshev / data-informed orthogonal polynomials

$$\text{Roundworm shape} = a_0(t) \text{ (straight line)} + a_1(t) \text{ (diagonal line)} + a_2(t) \text{ (U-shape)} + a_3(t) \text{ (S-shape)} + a_4(t) \text{ (W-shape)}$$

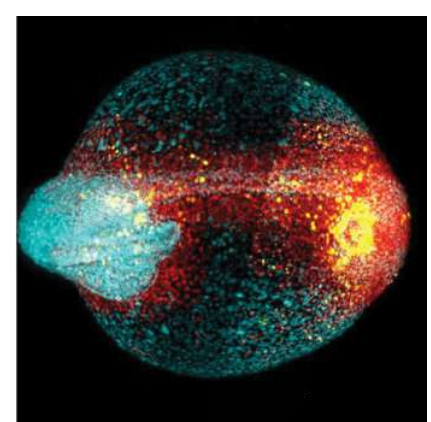
Jellyfish



Fourier-Jacobi basis functions

$$\text{Jellyfish shape} = a_{00}(t) \text{ (solid brown circle)} + a_{10}(t) \text{ (blue-to-orange gradient)} + a_{01}(t) \text{ (blue-to-orange vertical gradient)} + \dots + a_{33}(t) \text{ (complex pattern)}$$

Zebrafish



Scalar spherical harmonics

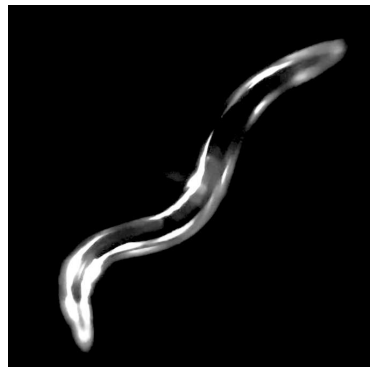
$$\text{Zebrafish shape} = a_{00}(t) \text{ (solid teal circle)} + a_{10}(t) \text{ (blue-to-yellow gradient)} + a_{11}(t) \text{ (blue-to-yellow vertical gradient)} + a_{20}(t) \text{ (blue-to-yellow horizontal gradient)} + a_{21}(t) \text{ (blue-to-yellow diagonal gradient)} + \dots$$

Vector spherical harmonics

$$b_{10}(t) \text{ (teal sphere with red arrows)} + b_{01}(t) \text{ (yellow sphere with red arrows)} + b_{20}(t) \text{ (blue sphere with red arrows)} + b_{02}(t) \text{ (blue sphere with red arrows)} + b_{30}(t) \text{ (blue sphere with red arrows)} + \dots$$

Can we learn predictive models of biophysical
dynamics in spectral space ?

Experimental motivation



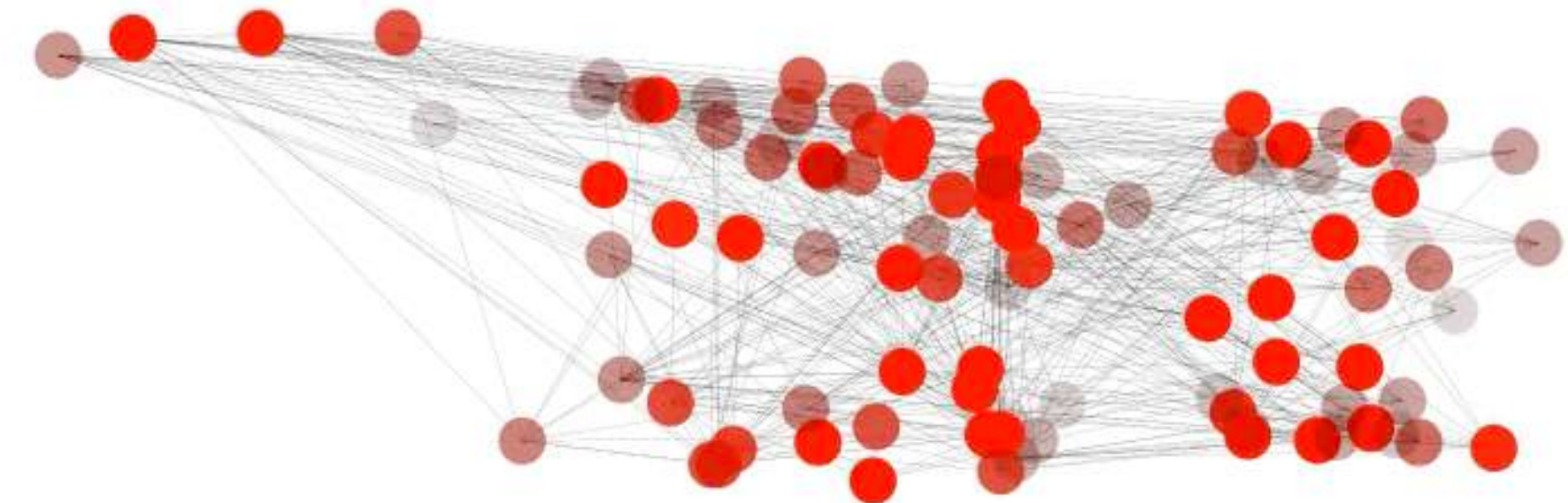
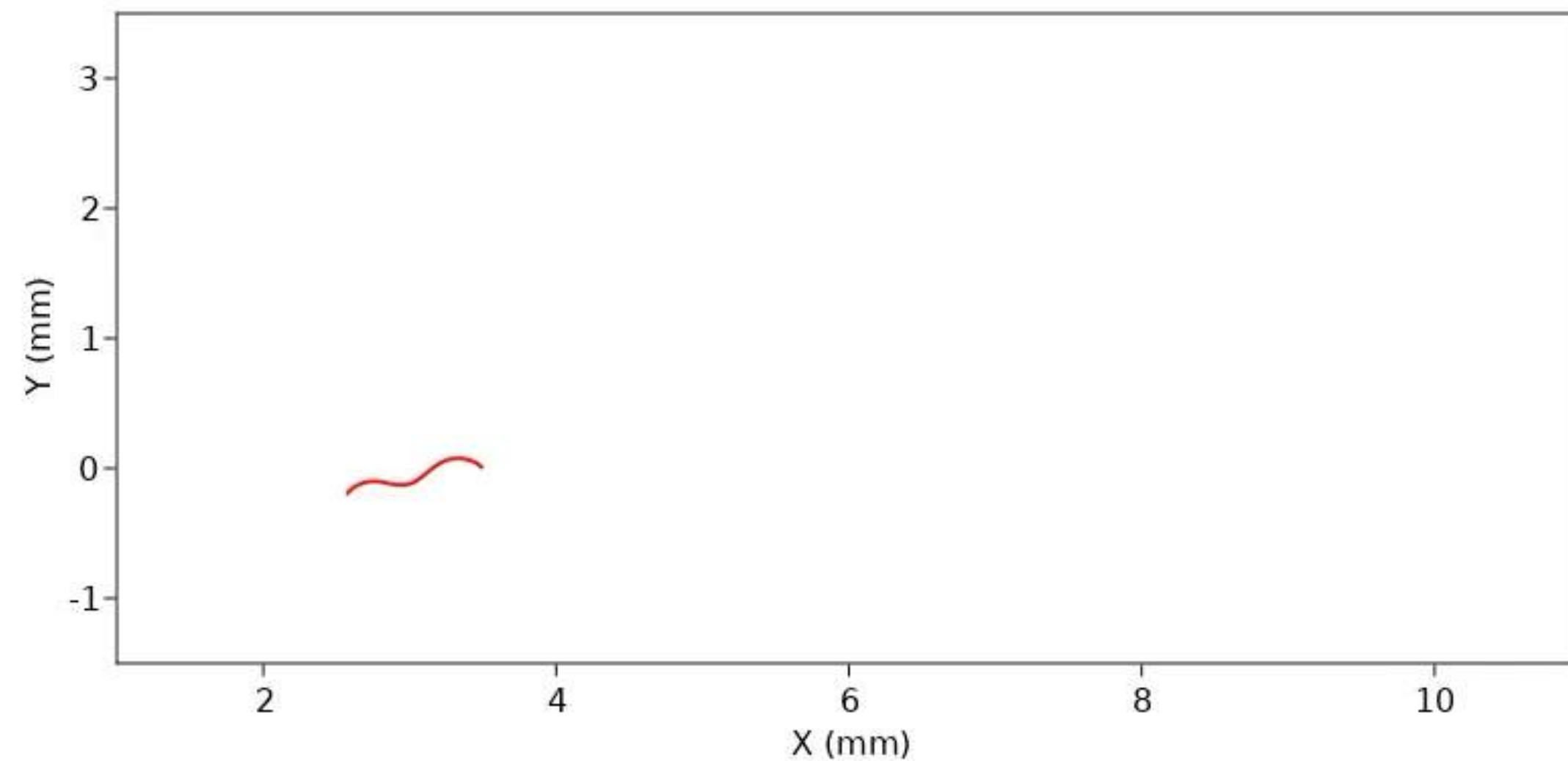
C. elegans

physical locomotion dynamics



spatiotemporal neuron activity

Phys Rev Lett 2023



Cohen & JD, preprint 2025



Steven Flavell



Neuromechanics

Model learning : increment-based

Given measured time series data $x(t_1), x(t_2), \dots$ find 'simple' dynamical model

Deterministic

$$\dot{x} = f(x | p)$$

$$0 = dx - f(x | p)dt$$

$$\min_p | dx - f(x | p)dt |$$

constraint optimization
(sparsity, symmetries, ...)

$$dx = x(t_{i+1}) - x(t_i)$$

$$dt = t_{i+1} - t_i$$

Model learning : increment-based

Given measured time series data $x(t_1), x(t_2), \dots$ find 'simple' dynamical model

Deterministic

$$\dot{x} = f(x | p)$$

$$0 = dx - f(x | p)dt$$

$$\min_p | dx - f(x | p)dt |$$

constraint optimization
(sparsity, symmetries, ...)

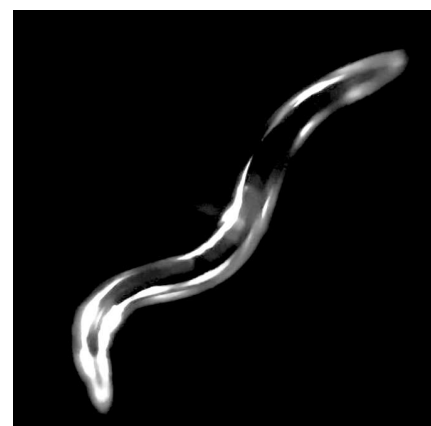
Stochastic

$$dx(t) = f(x | p)dt + g(x | p)dB(t)$$

$$dB = g^{-1}(dx - f dt)$$

$$\mathbb{P}[dB] = \mathbb{P}[g^{-1}(dx - f dt)]$$

maximum likelihood estimation of p

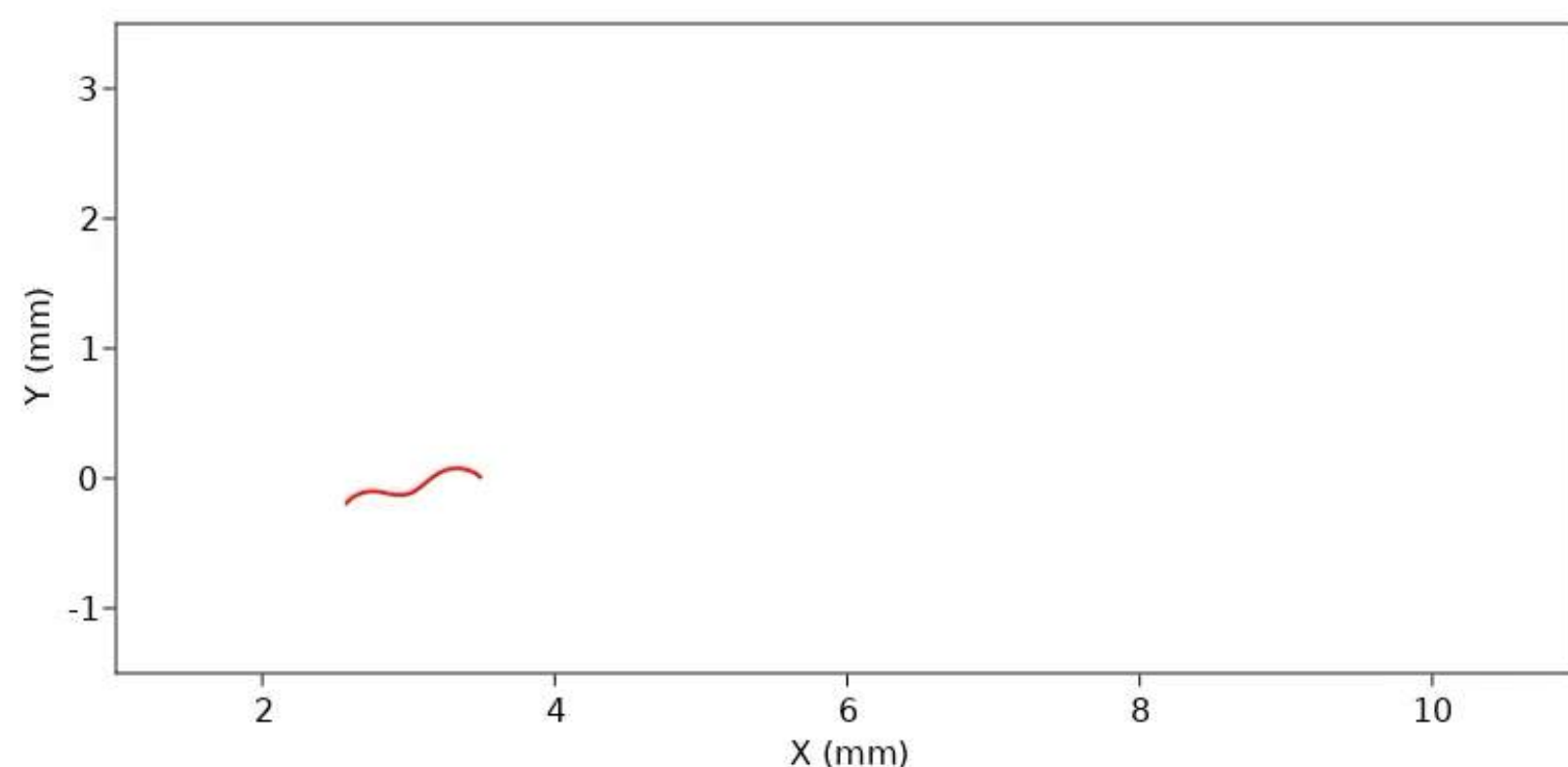


Learning linear models for undulatory locomotion dynamics



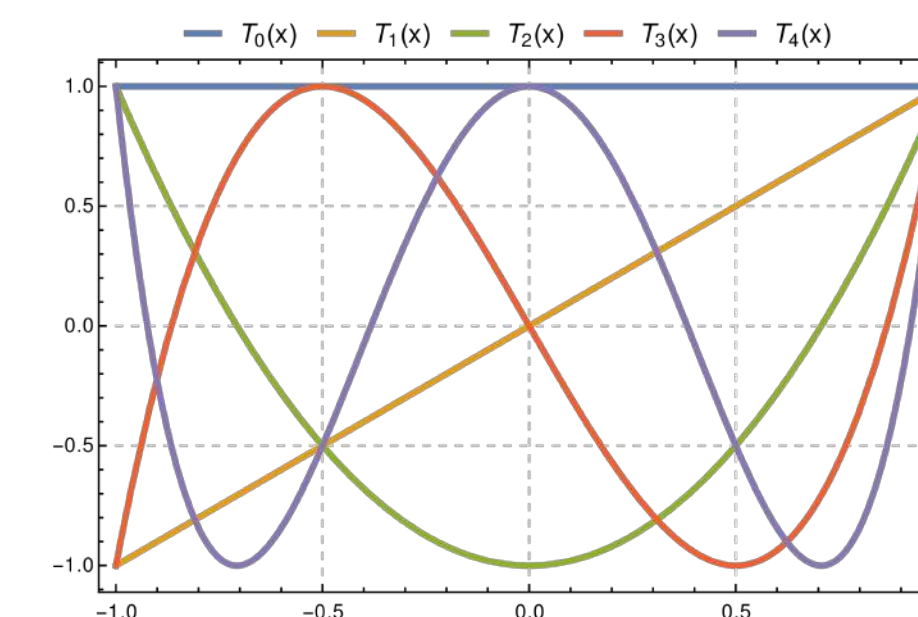
Flavell lab
(MIT BCS)

Experimental worm trajectories



$$\begin{bmatrix} x(s, t) \\ y(s, t) \end{bmatrix} = \sum_{k=0}^n T_k(s) \begin{bmatrix} \hat{x}_k(t) \\ \hat{y}_k(t) \end{bmatrix}$$

$$\begin{bmatrix} \hat{x}_k(t) \\ \hat{y}_k(t) \end{bmatrix} = \frac{\gamma_n}{\pi} \int_{-1}^1 ds \, w(s) T_k(s) \begin{bmatrix} x(s, t) \\ y(s, t) \end{bmatrix}$$



Linear short-time dynamics

$$\frac{d}{dt} \begin{bmatrix} \hat{\mathbf{x}} \\ \hat{\mathbf{y}} \end{bmatrix} = \mathbf{M} \begin{bmatrix} \hat{\mathbf{x}} \\ \hat{\mathbf{y}} \end{bmatrix}$$

$$\psi(t) \sim \begin{bmatrix} \hat{x}_1 + i\hat{y}_1 \\ \vdots \\ \hat{x}_N + i\hat{y}_N \end{bmatrix}$$

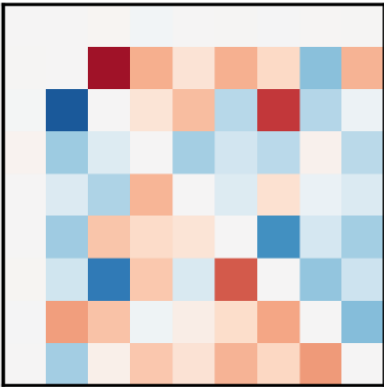
Later: complex representation

Symmetry & biophysical constraints

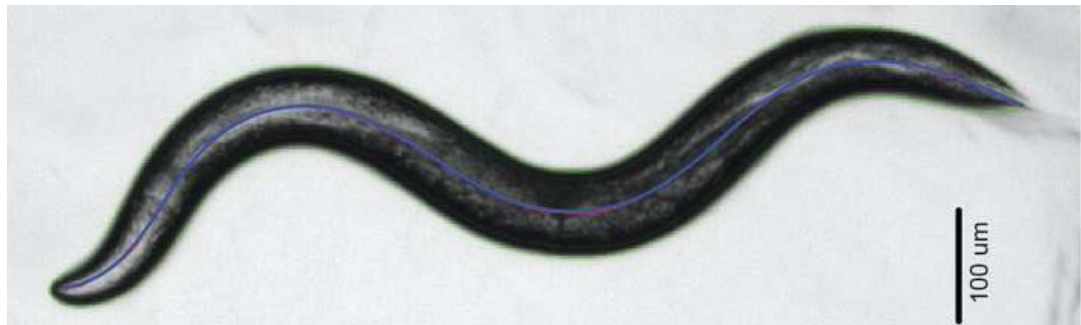
Constraints	Model (straight motion)	Degrees of freedom	
None	$\frac{d}{dt} \begin{bmatrix} \hat{\mathbf{x}} \\ \hat{\mathbf{y}} \end{bmatrix} = \mathbf{M}^{(0)} \begin{bmatrix} \hat{\mathbf{x}} \\ \hat{\mathbf{y}} \end{bmatrix}$	$2N \times 2N$	☹
Rotational Invariance	$\frac{d}{dt} \begin{bmatrix} \hat{\mathbf{x}} \\ \hat{\mathbf{y}} \end{bmatrix} = \begin{bmatrix} \mathbf{M}^{(1)} & 0 \\ 0 & \mathbf{M}^{(1)} \end{bmatrix} \begin{bmatrix} \hat{\mathbf{x}} \\ \hat{\mathbf{y}} \end{bmatrix}$	$N \times N$	☹
Length approx conserved	$\frac{d}{dt} \begin{bmatrix} \hat{\mathbf{x}} \\ \hat{\mathbf{y}} \end{bmatrix} = \begin{bmatrix} \mathbf{M}^{(2)} \mathbf{W} & 0 \\ 0 & \mathbf{M}^{(2)} \mathbf{W} \end{bmatrix} \begin{bmatrix} \hat{\mathbf{x}} \\ \hat{\mathbf{y}} \end{bmatrix}$	$\frac{N \times (N - 1)}{2}$	😊

$$\mathbf{W}_{ij} = \int_{-1}^1 ds \left(\frac{dT_i}{ds} \frac{dT_j}{ds} \right)$$

$$\mathbf{M}^{(2)} = -\mathbf{M}^{(2)\top}$$

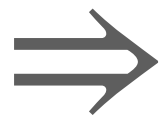


Flavell lab (MIT Picower)



Planar curve

$$x(s, t) + iy(s, t)$$

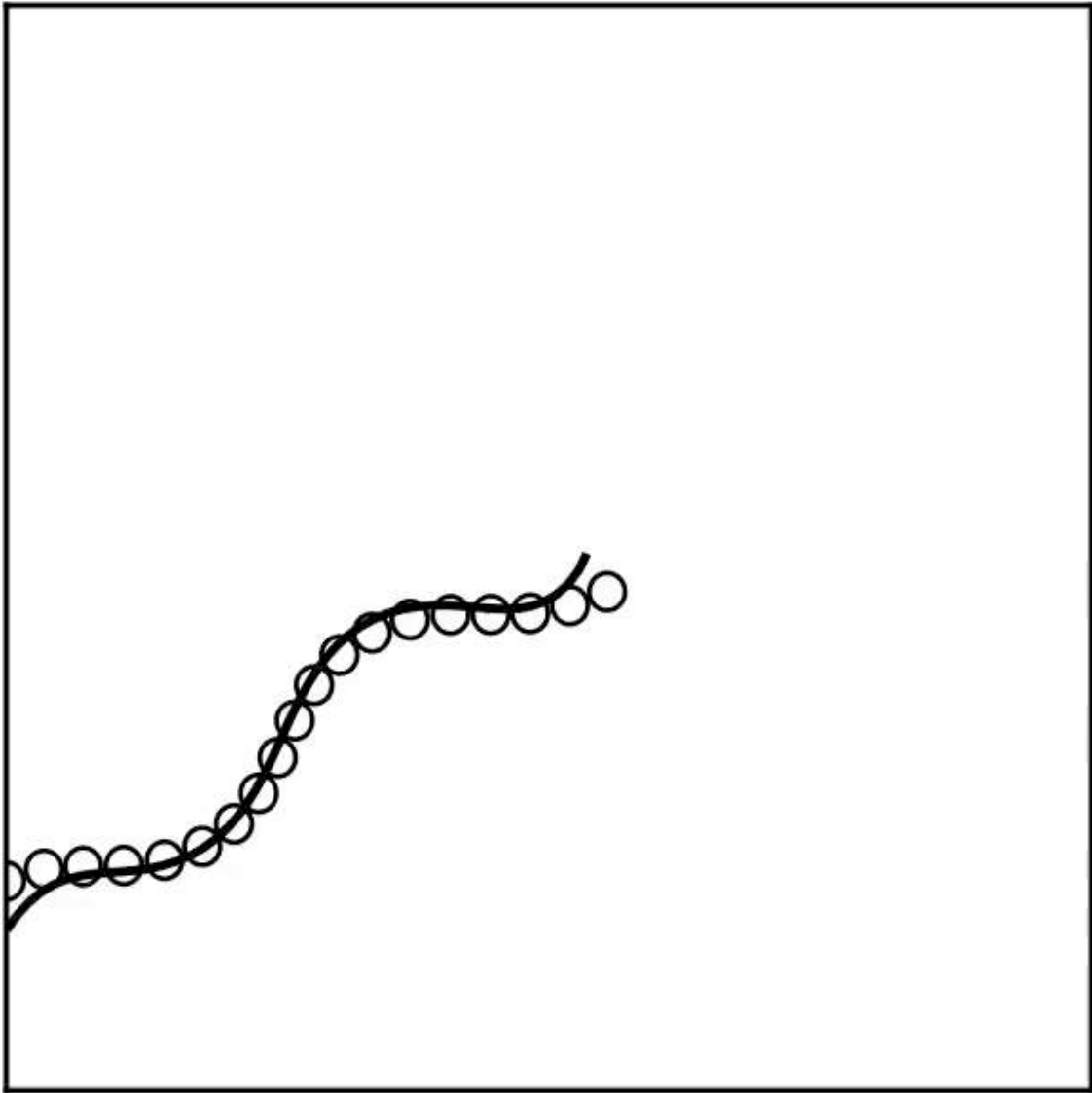


Center-of-mass

$$\psi_0(t) \sim \hat{x}_0 + i\hat{y}_0$$

Shape modes

$$\psi(t) \sim \begin{bmatrix} \hat{x}_1 + i\hat{y}_1 \\ \vdots \\ \hat{x}_N + i\hat{y}_N \end{bmatrix}$$



$$\psi^\dagger \psi = 1$$

‘Length’ constraint

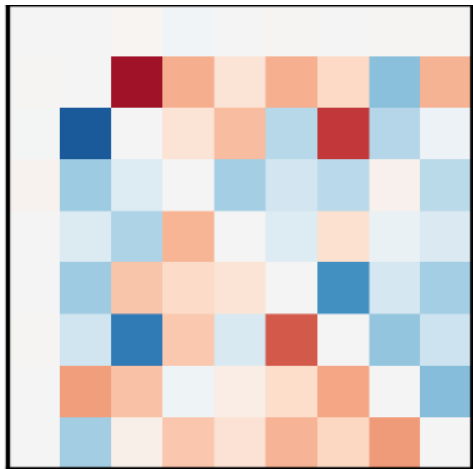
$$i\dot{\psi} = H\psi$$

Shape mode dynamics

$$\dot{\psi}_0 = H_0\psi$$

Propulsion dynamics

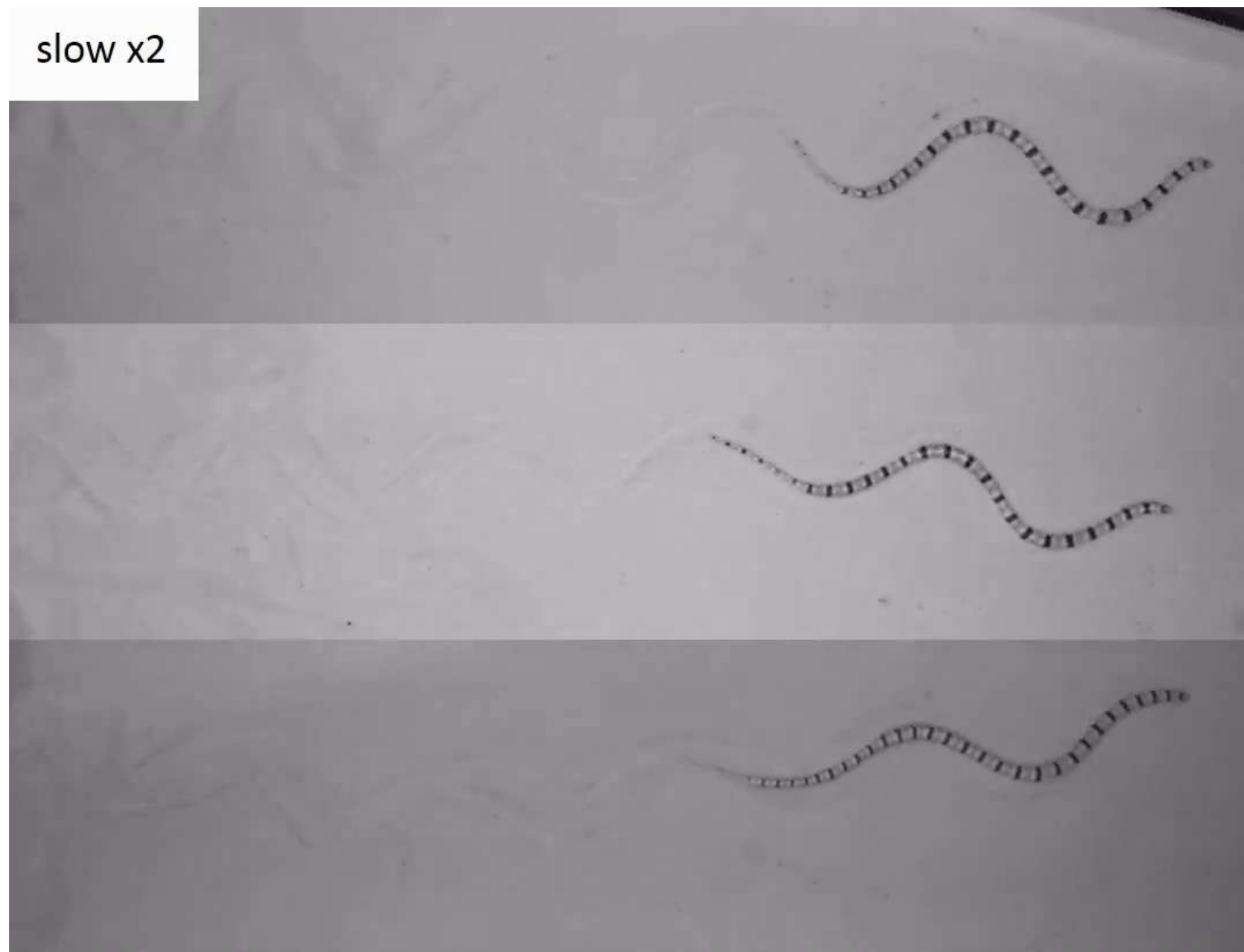
H



$$H = \cancel{S} + iA$$

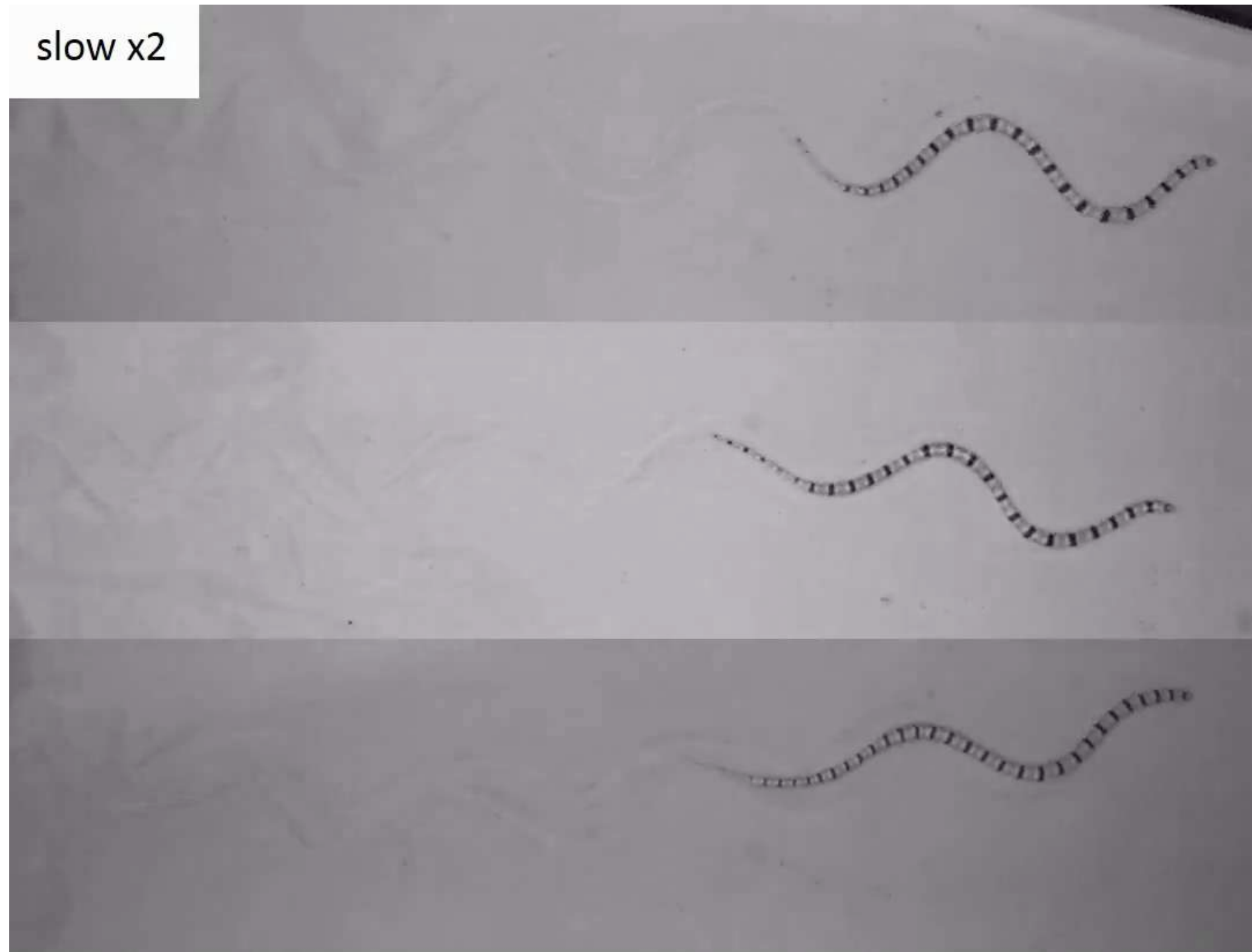
Straight motion

Desert snake



Goldman Lab (Georgia Tech), PNAS 2019

Desert snake



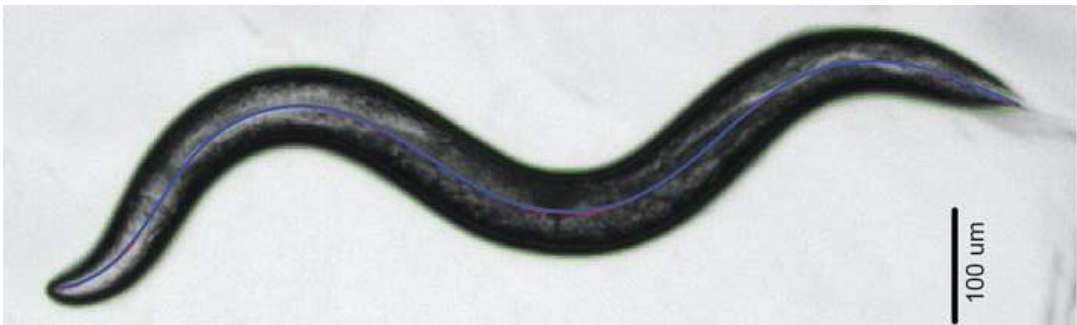
Goldman Lab (Georgia Tech), PNAS 2019

Toy snake



Cohen*, Hastewell* et al, PRL 2023

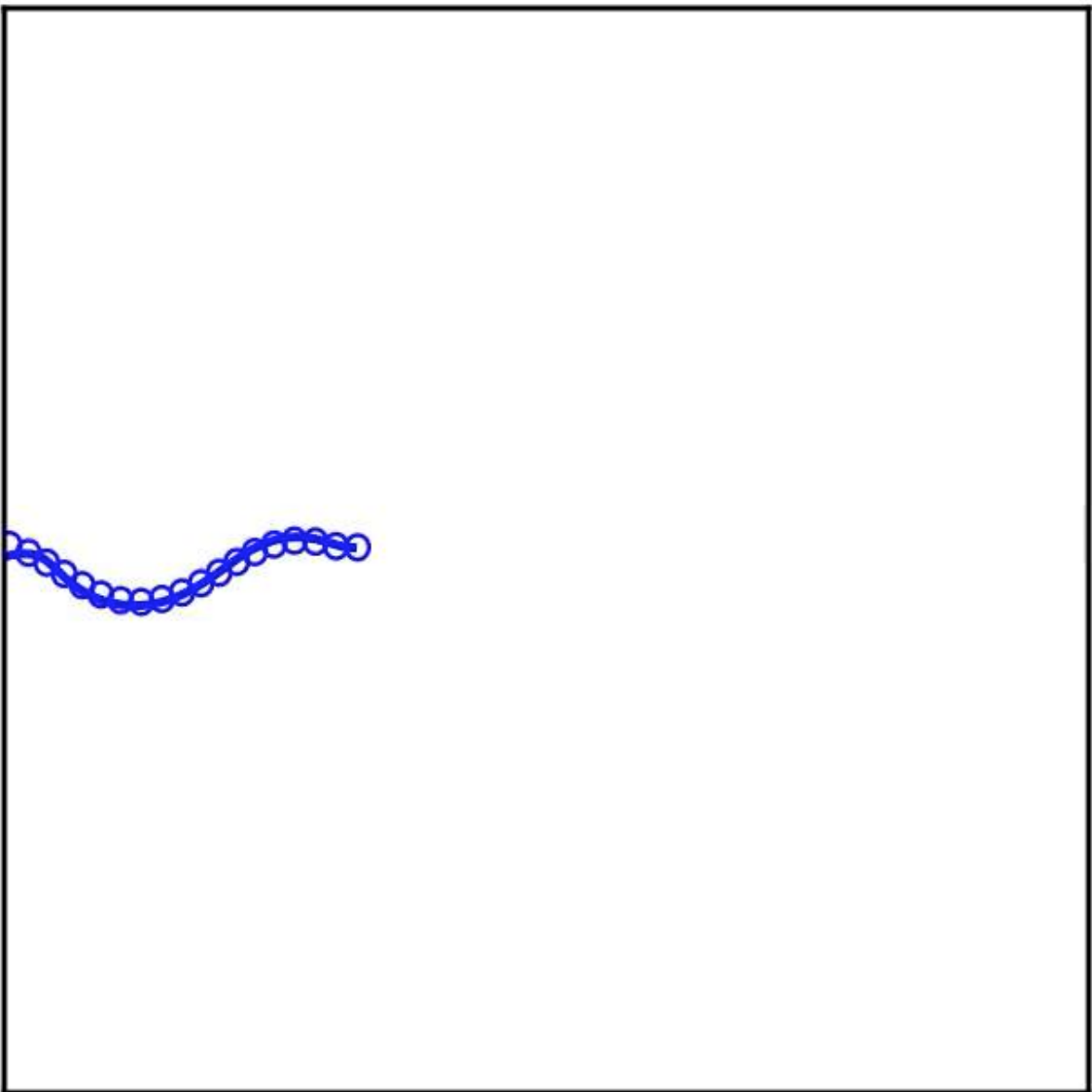
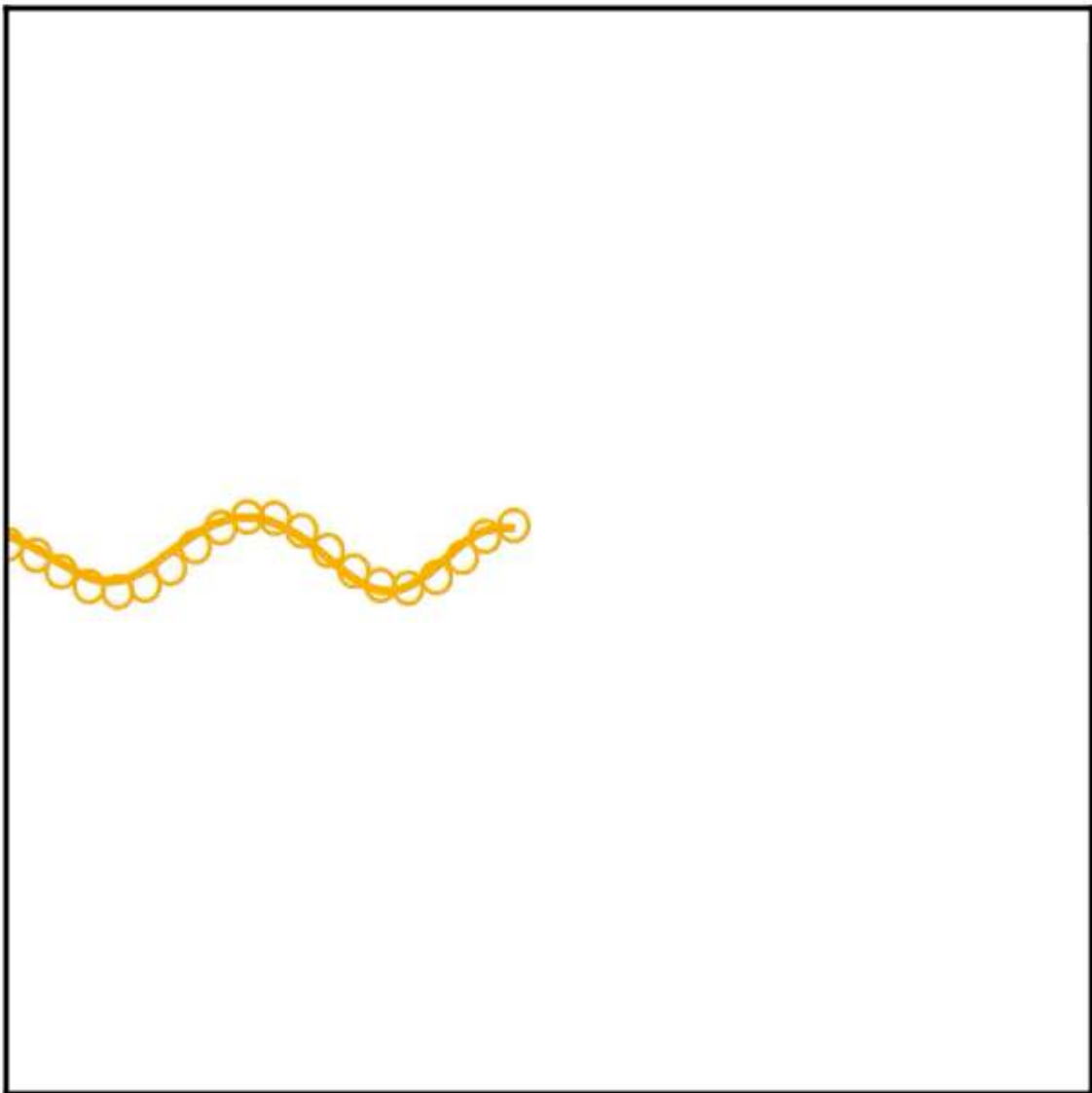
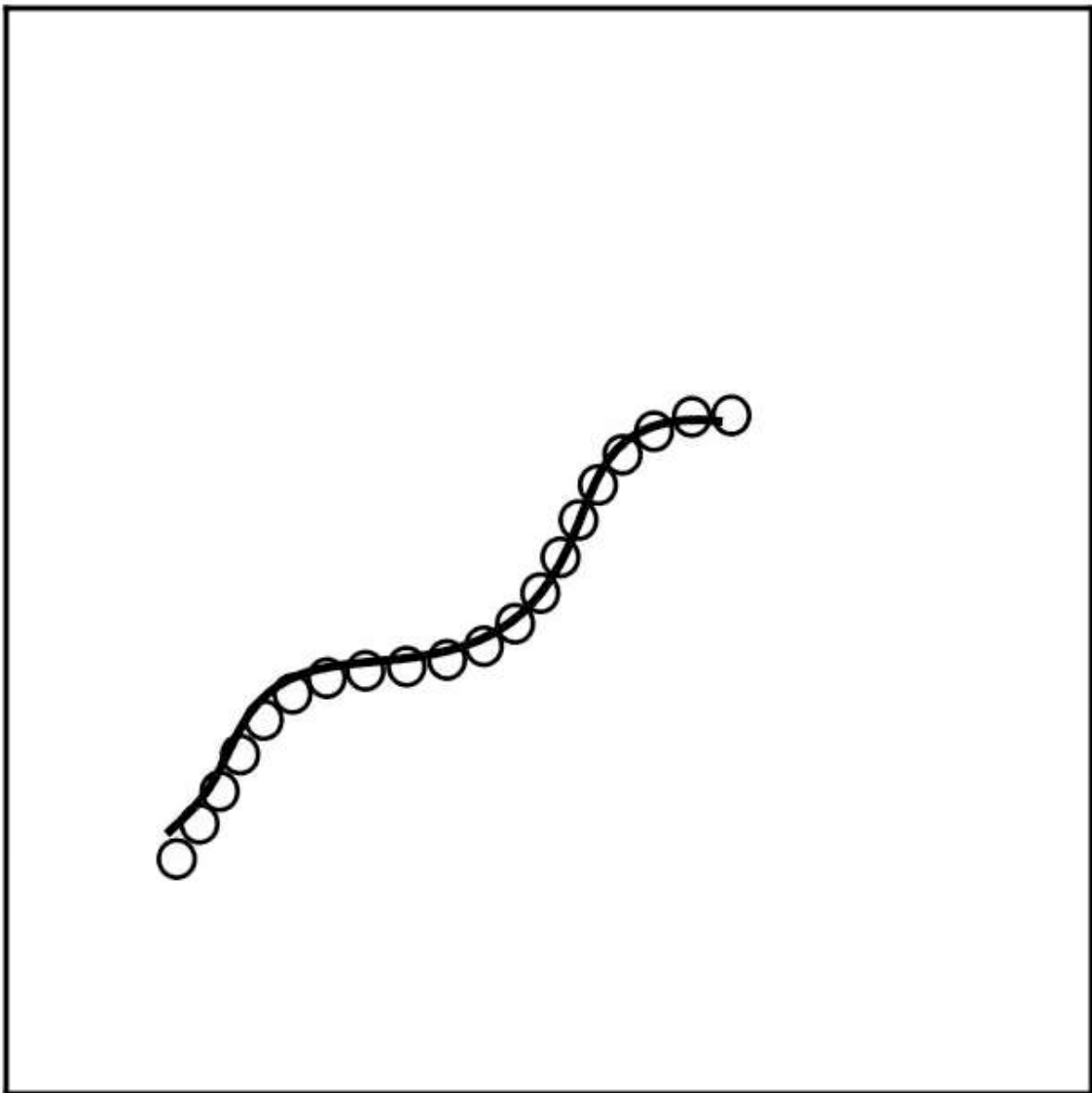
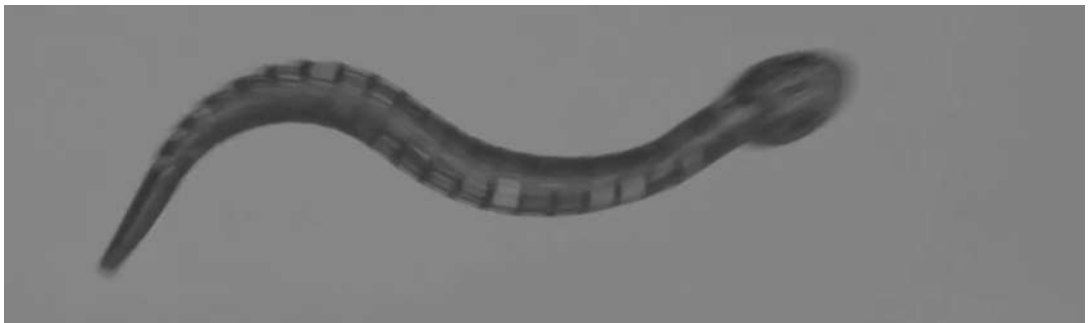
Flavell lab (MIT Picower)



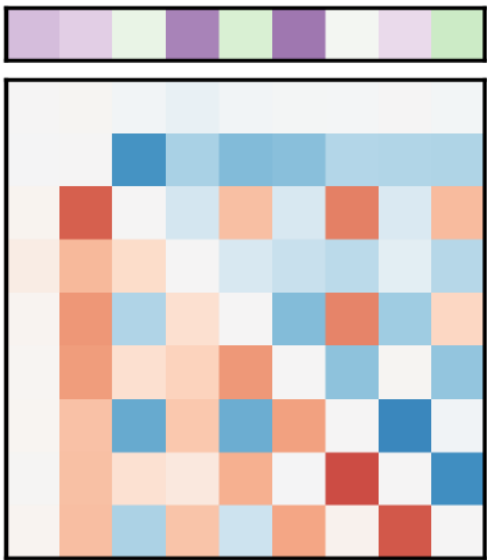
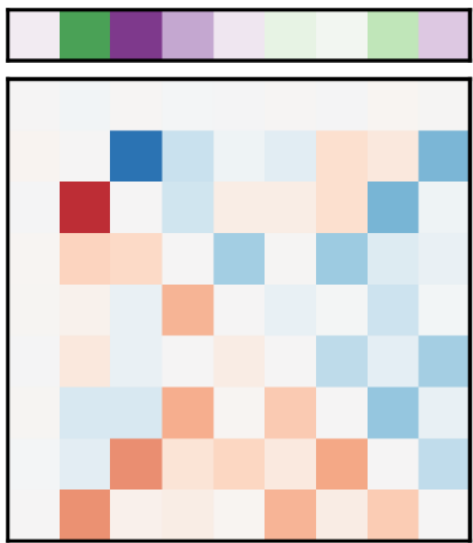
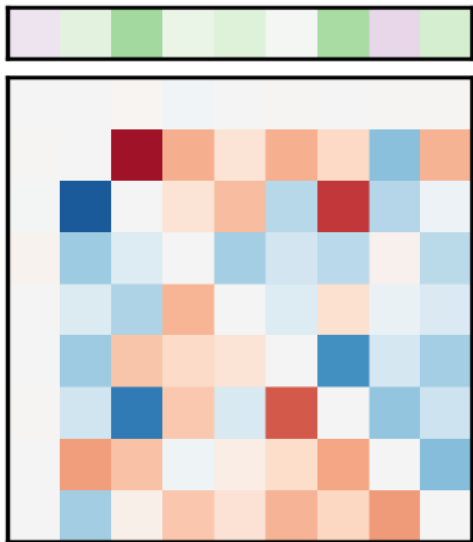
Goldman lab (Georgia Tech)



Alex's dorm room (MIT)



H



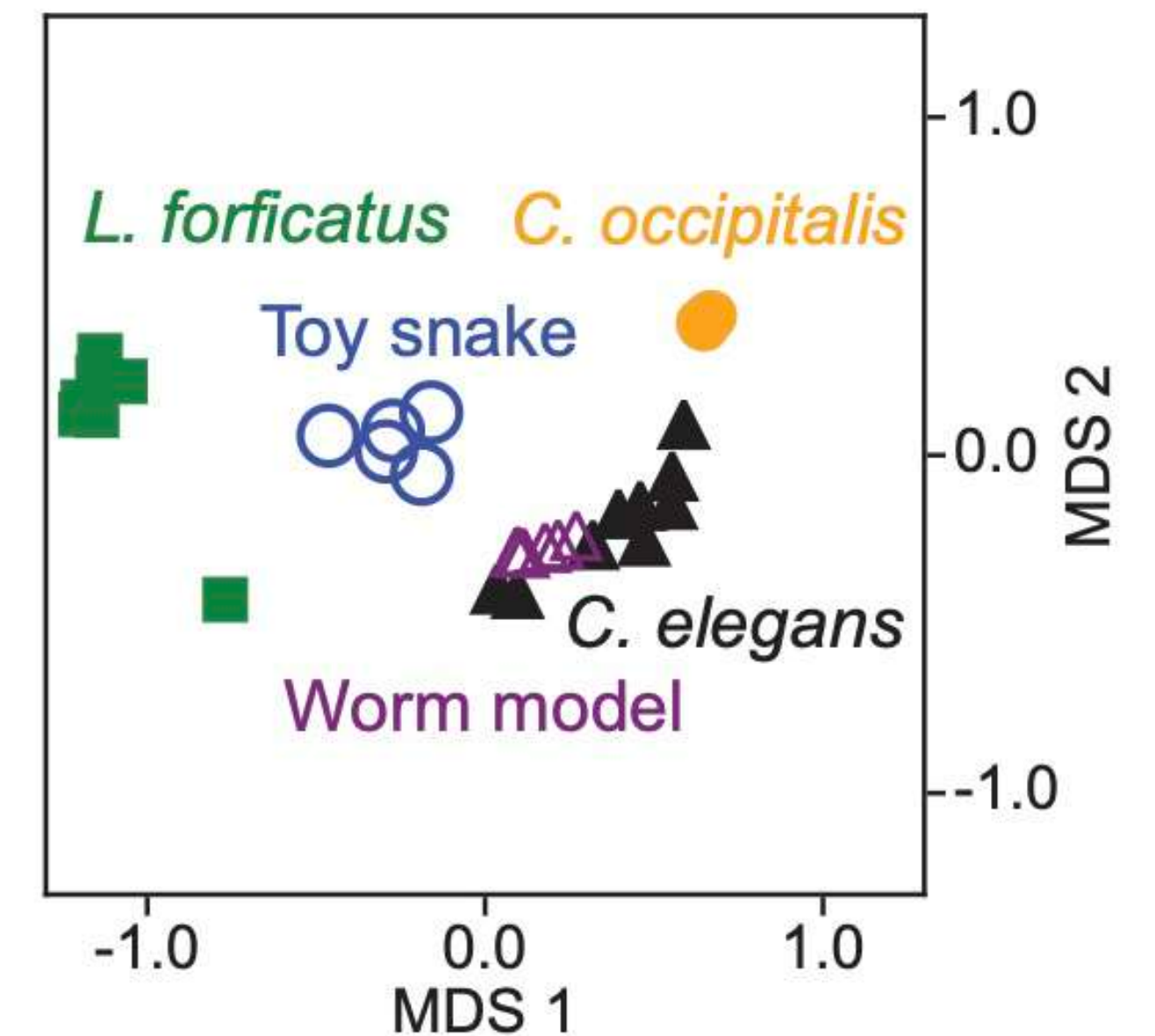
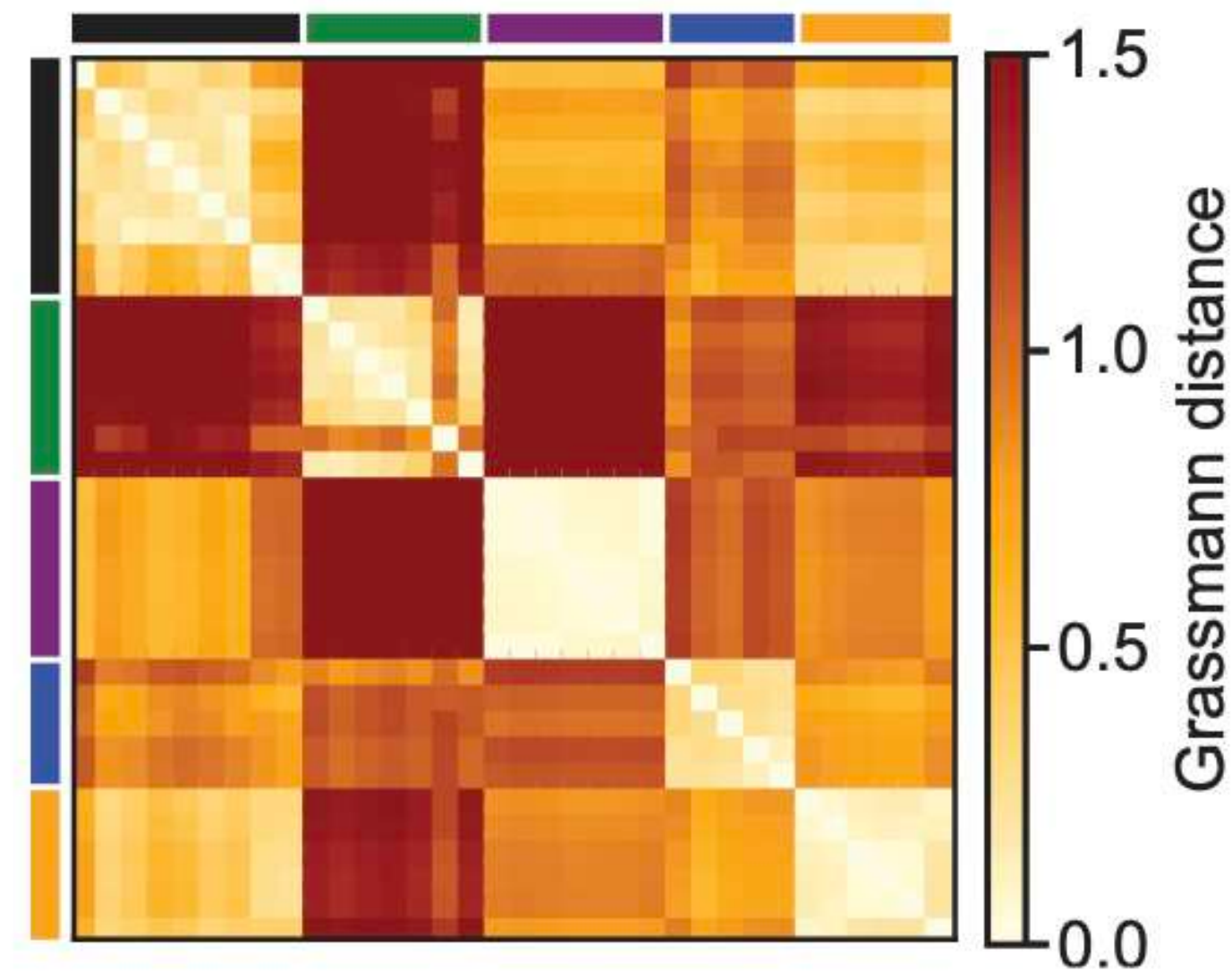
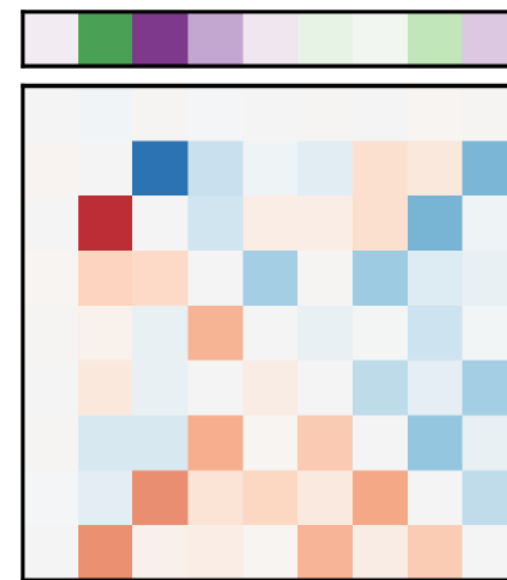
Schrödinger dynamics of undulatory locomotion

$$\begin{bmatrix} x(s, t) \\ y(s, t) \end{bmatrix} = \sum_{k=0}^n T_k(s) \begin{bmatrix} \hat{x}_k(t) \\ \hat{y}_k(t) \end{bmatrix}$$

$$\dot{\psi}_0 = H_0 \psi$$

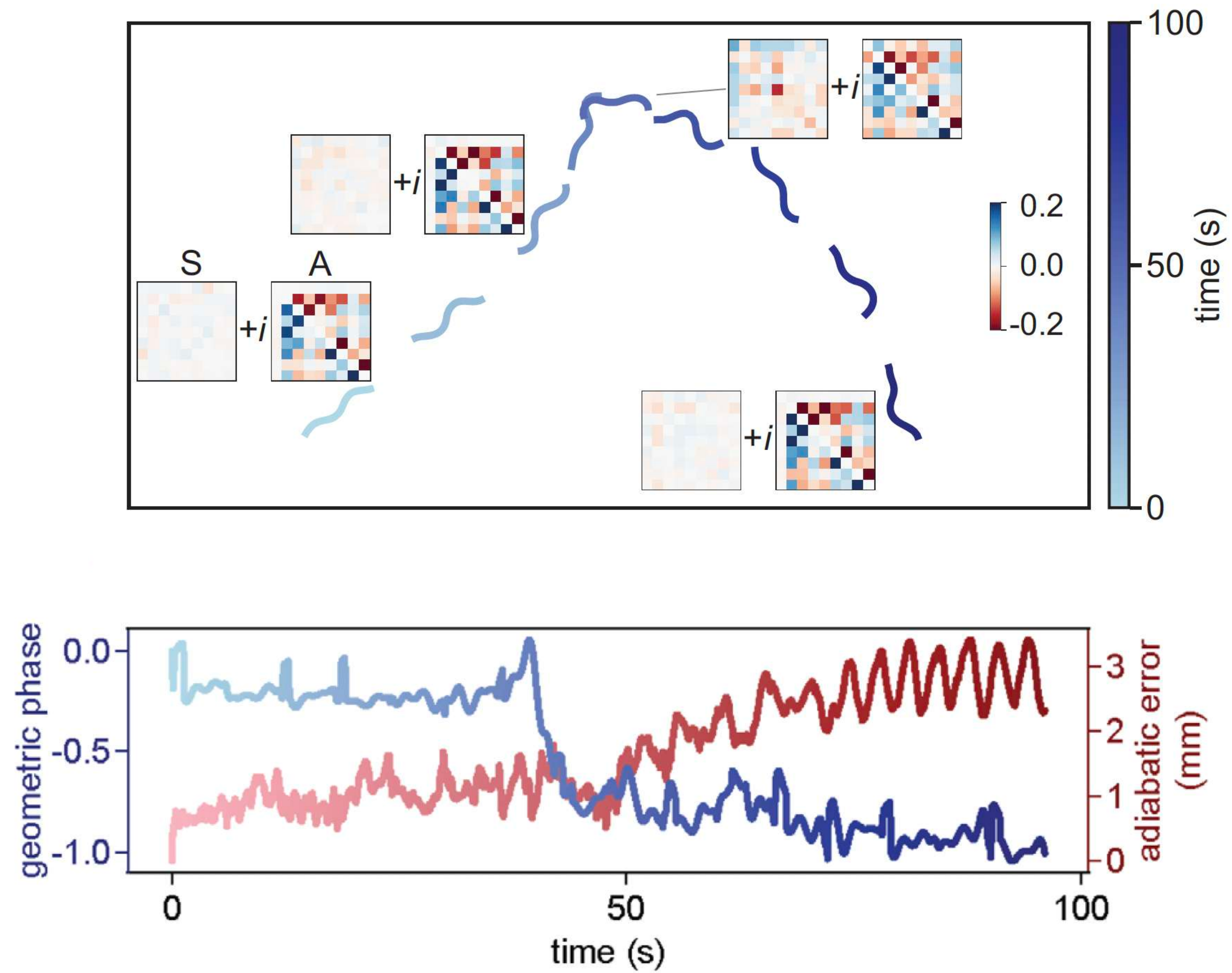
$$i\dot{\psi} = H\psi$$

$$1 = \psi^\dagger \psi$$

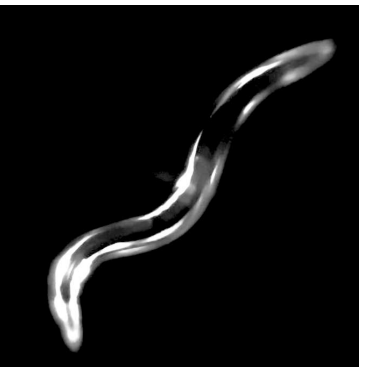


behavioral state $\Leftrightarrow$ low-rank Hamiltonian $\Rightarrow$ interspecies comparison

Time-dependent Hamiltonian and non-adiabatic turning



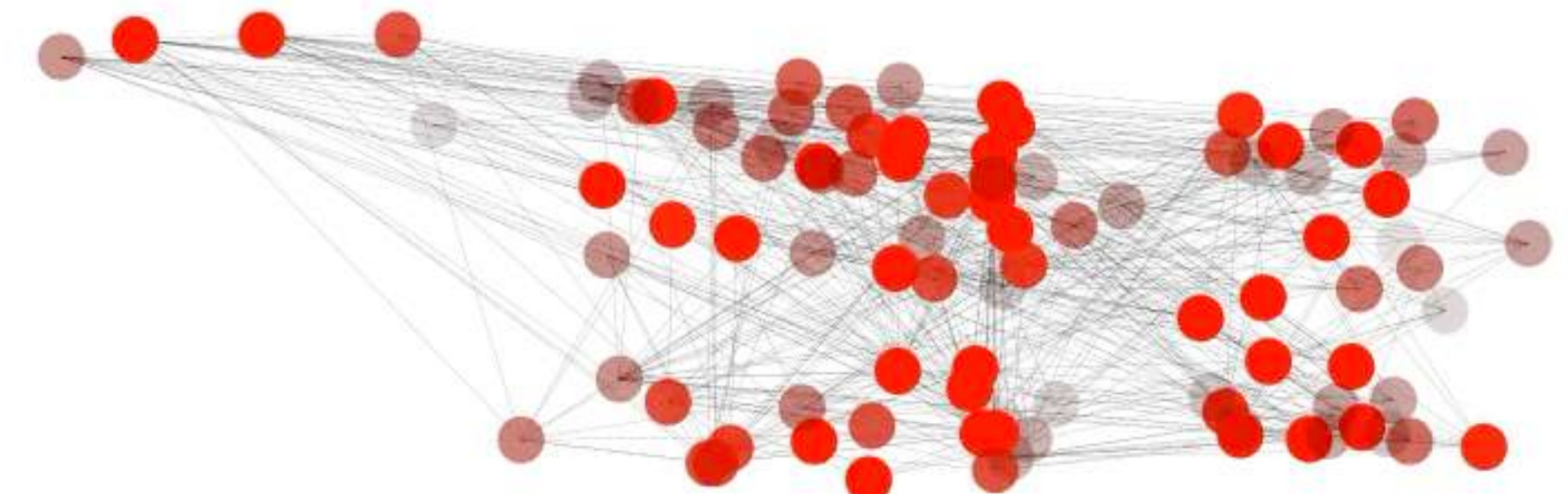
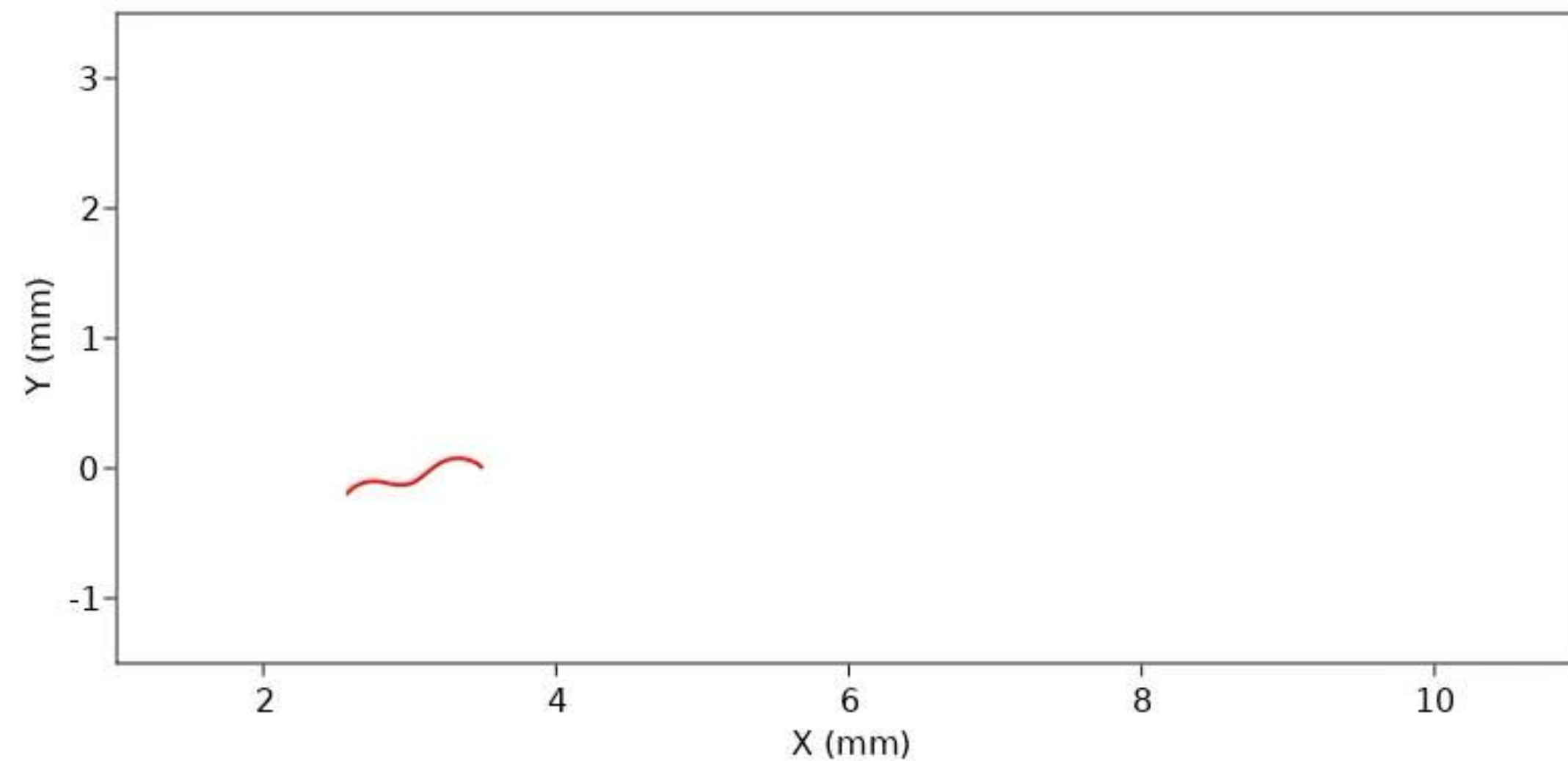
Outlook : Nonlinear model inference



C. elegans

Spatiotemporal neuron activity $\Leftrightarrow$ physical dynamics

Phys Rev Lett 2023



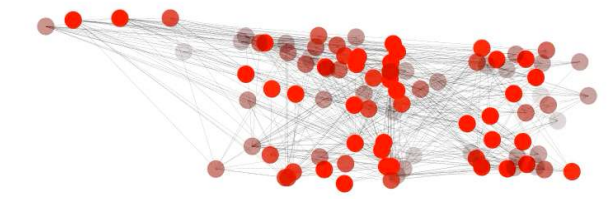
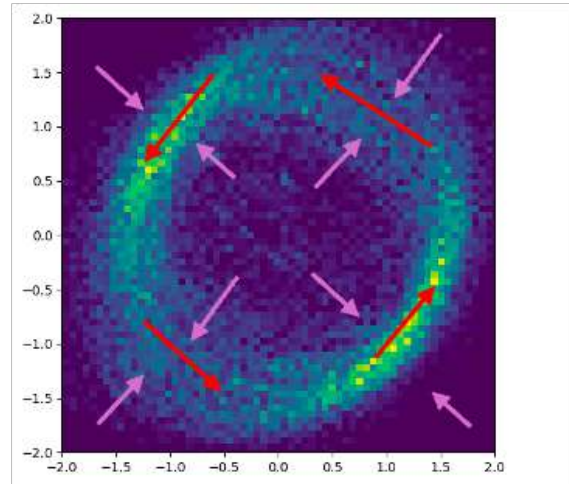
Steven Flavell



Neuromechanics

Cohen et al, preprint 2025 (available on request)

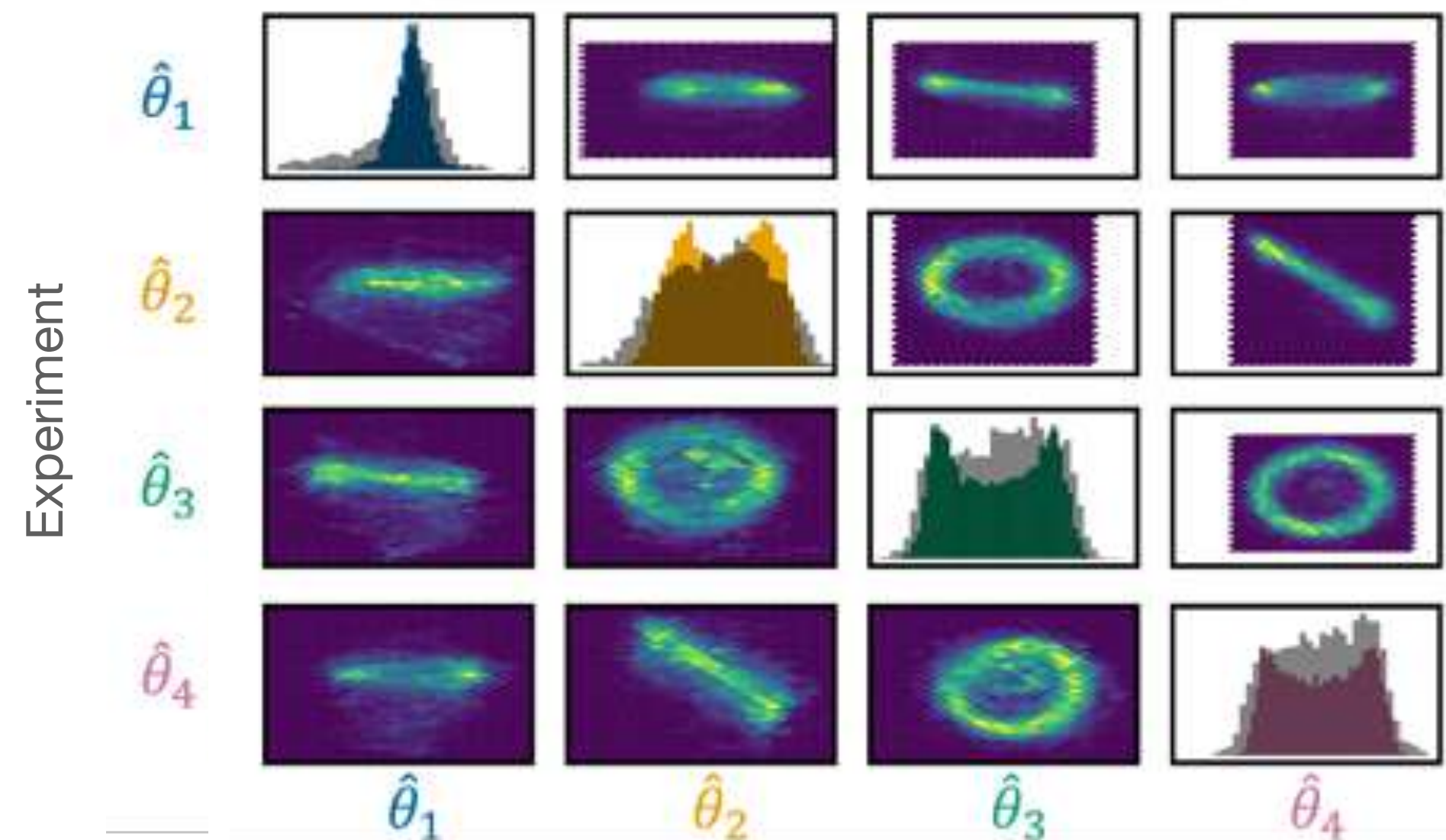
Outlook : Nonlinear SDE inference



$$d\hat{\theta}_i = f_i(\hat{\theta} | \mathbf{z}) + \sigma_{ij}(\mathbf{z}) dB_j(t)$$

$$\mathbf{f} = \nabla V + \mathbf{c}$$

Nonlinear SDE model

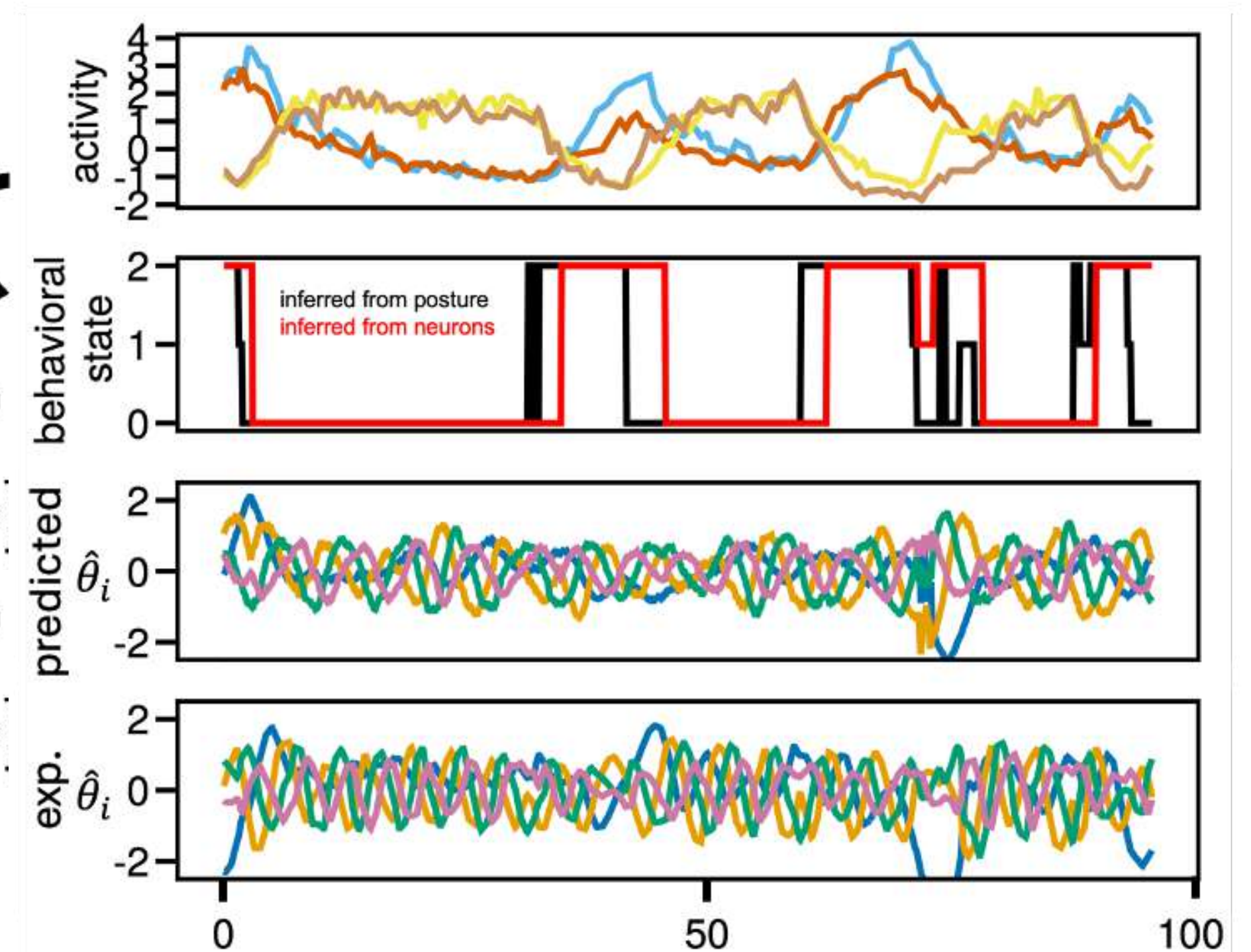


Predicting shape-dynamics from neuron activity

infer behavior
from neural
activity

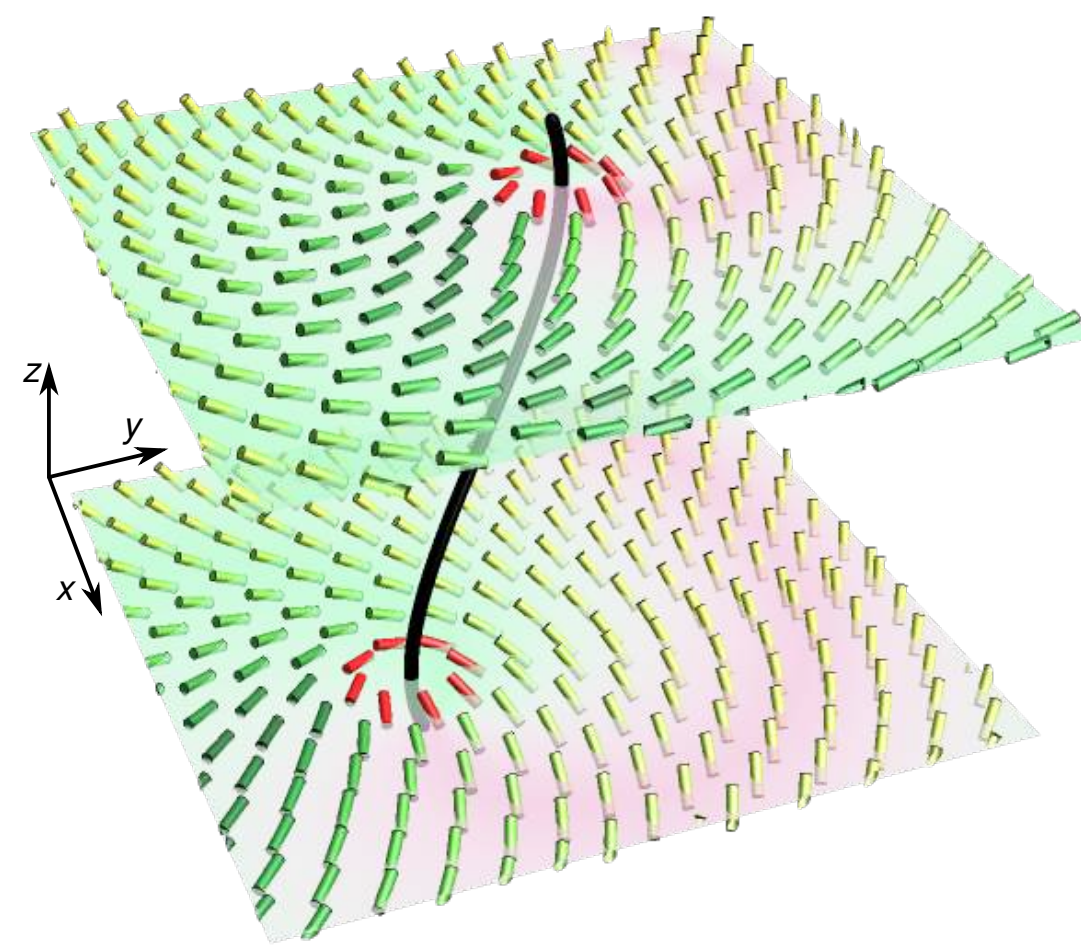
prediction
motion from
behavior

compare with
measured
motion



Part 2:

Topological defects & computation in liquid crystals



Science Advances 2022



PRX 2022

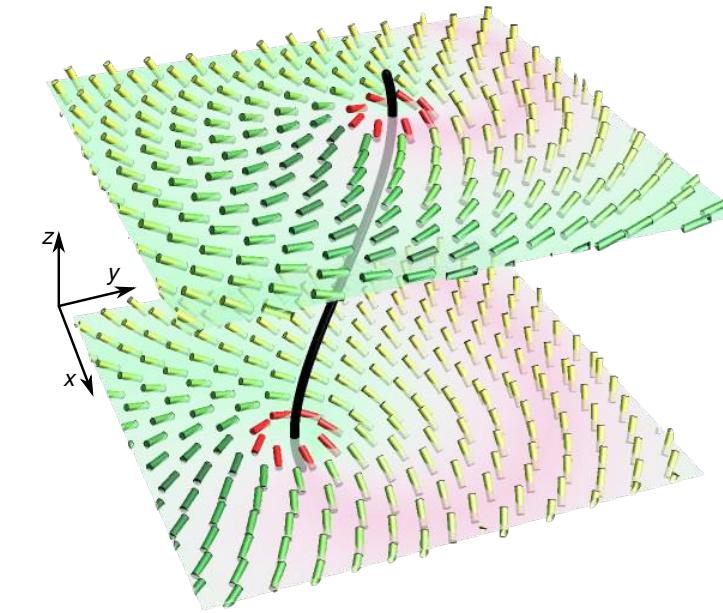
Broader motivation

Can we use topological defects as computational ‘bits’ ?

What types of computations can we perform ?

How can we manipulate defects to implement certain logical operations ?

Borrow ideas from other fields by exploring analogies ?



Science Advances 2022



Ziga Kos
(MIT $\Rightarrow$ Ljubljana)



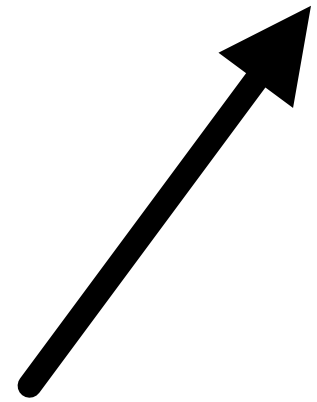
PRX 2022



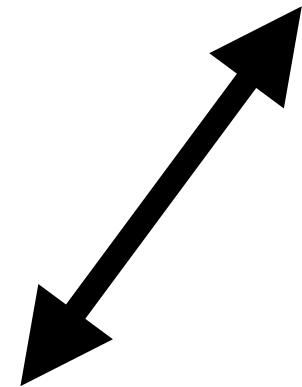
Alex Mietke
(MIT $\Rightarrow$ Oxford)

Dihedral particle symmetry

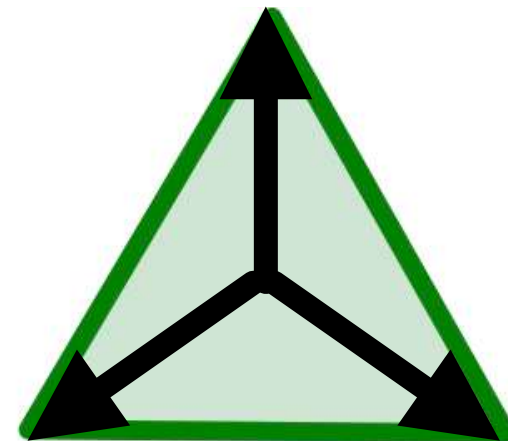
$k = 1$



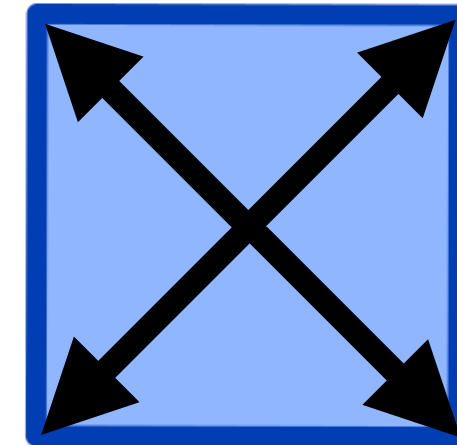
$k = 2$



$k = 3$



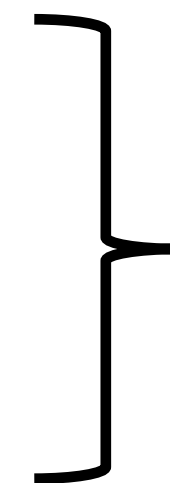
$k = 4$



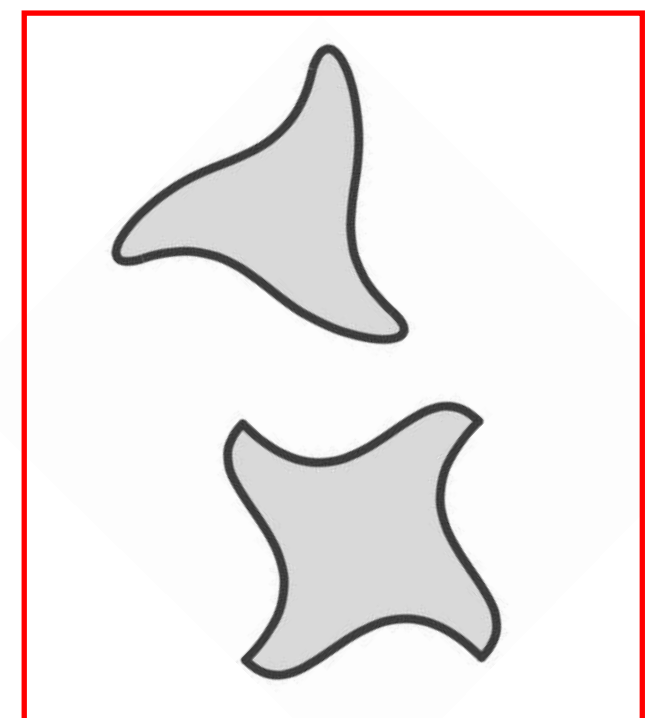
etc.

Symmetric under $\frac{2\pi}{k}$ rotation

~~+ reflection symmetries~~

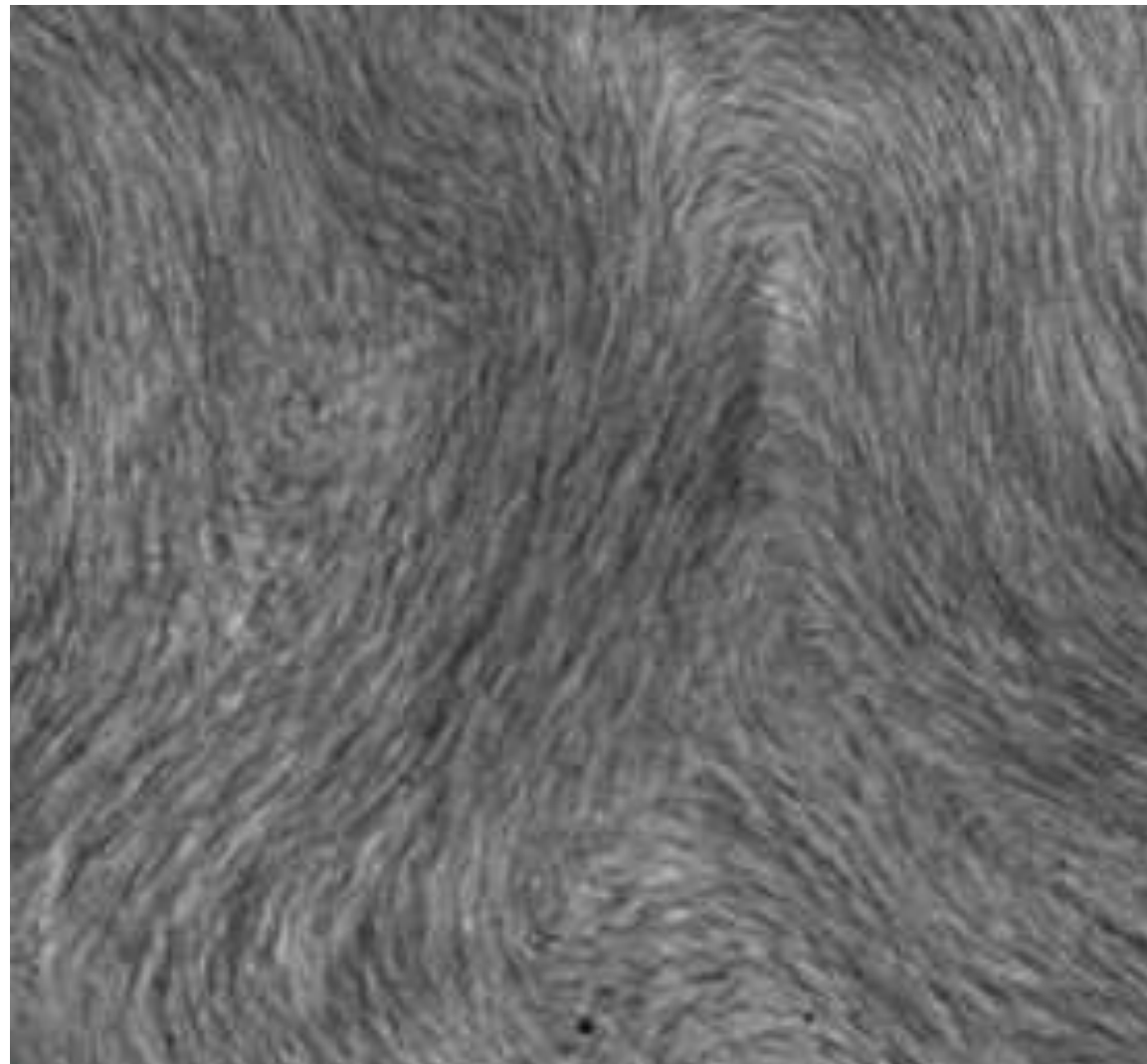


~~Cyclic group C_n~~
Dihedral group D_n



Two-dimensional dihedral ('k-atic') liquid crystals

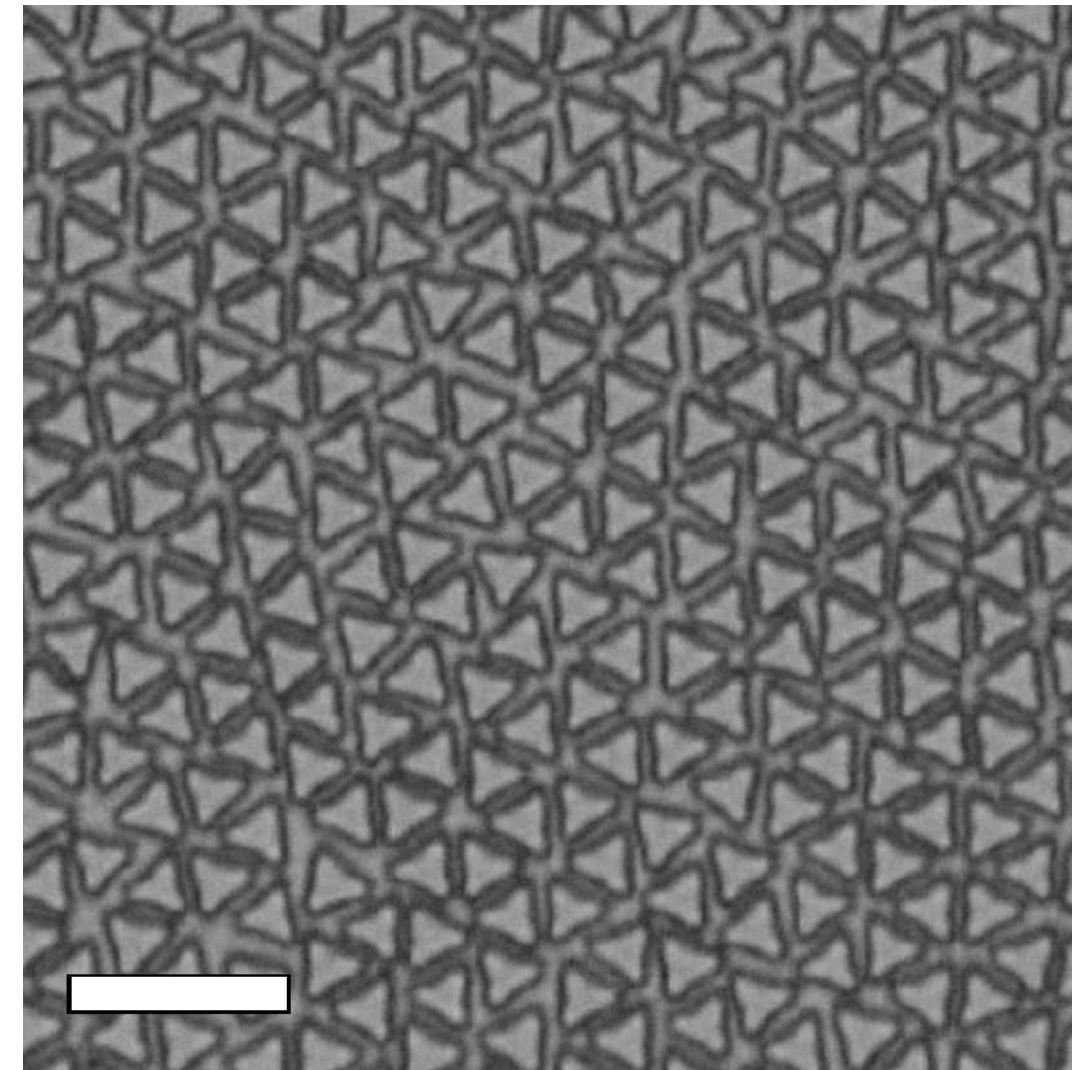
2-atic
(nematic)



Zhou et al. PNAS 2014

bacteria

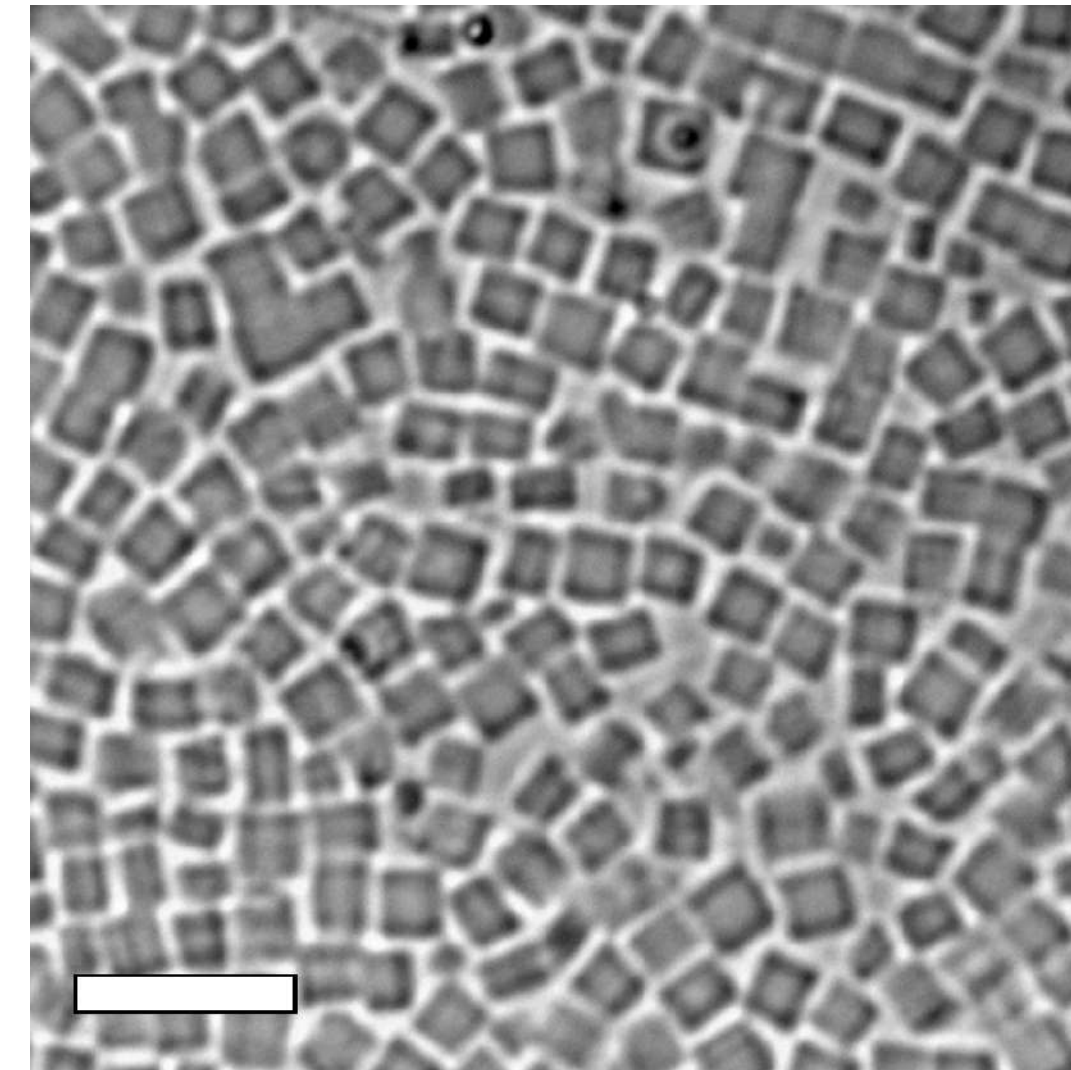
3-atic



Zhao et al. Nat Comm 2012

colloids

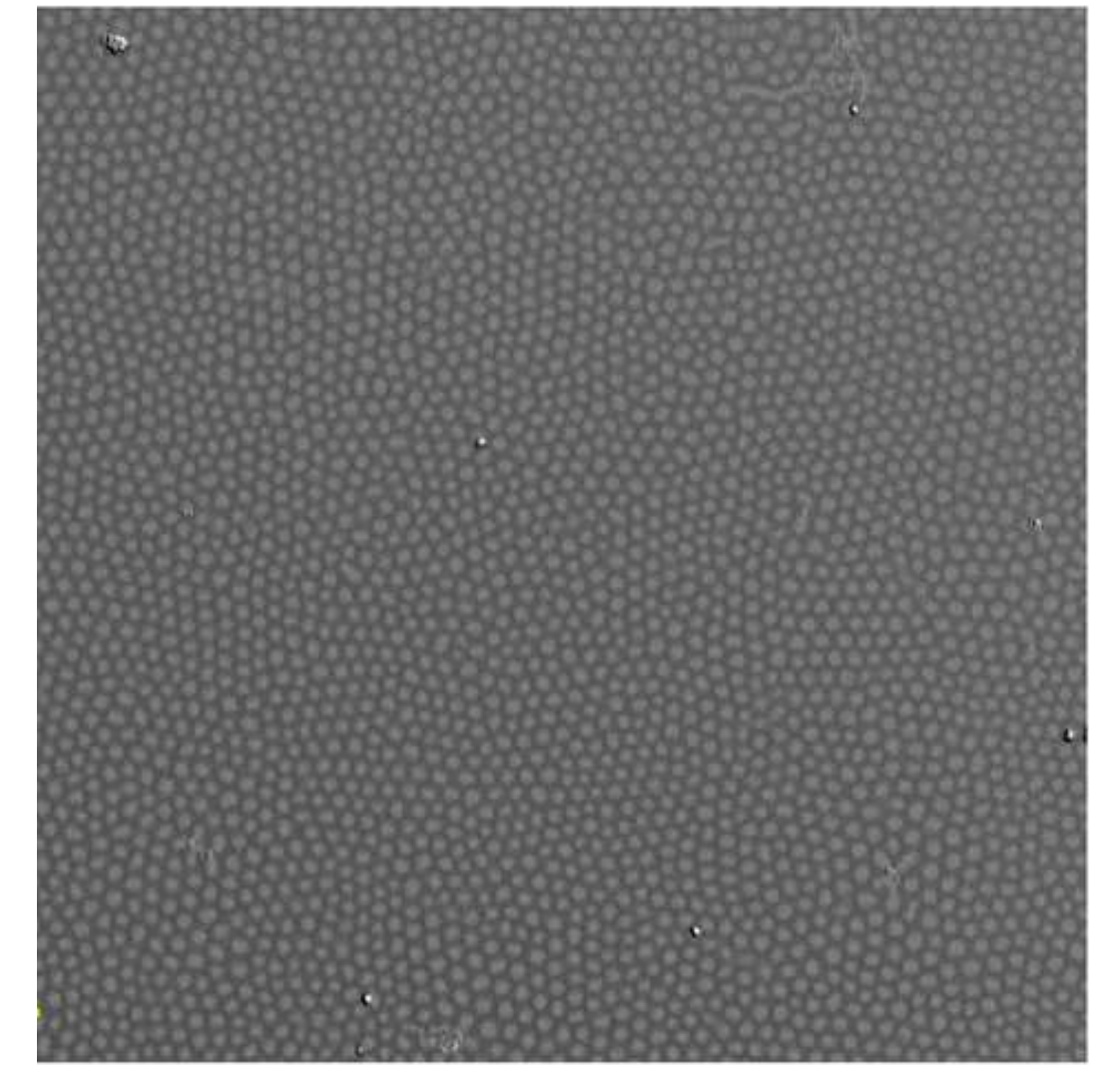
4-atic



Löffler. Thesis 2018 (Konstanz)

colloids

6-atic



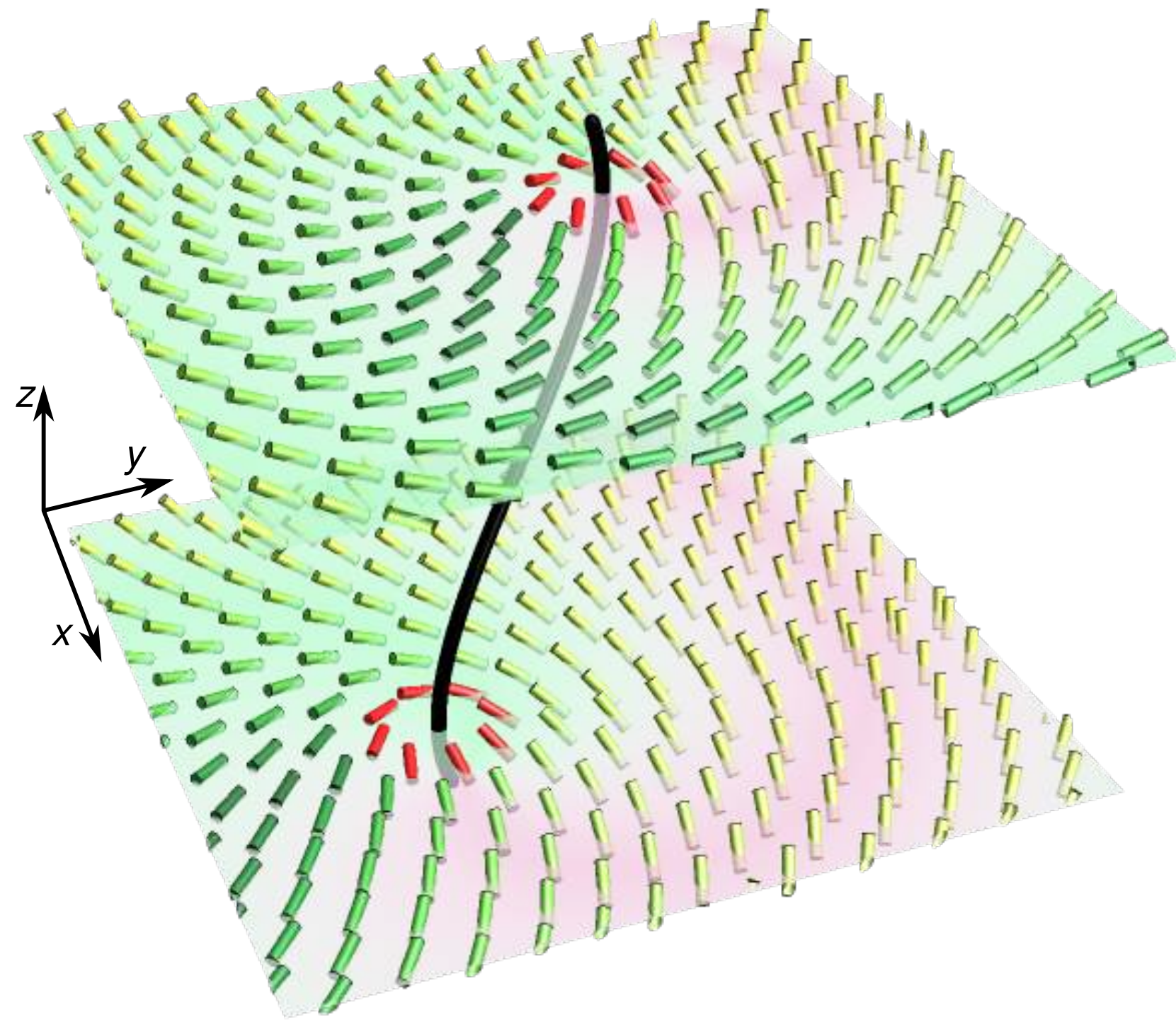
Zazvorka et al. Adv Funct Mater 2020

skyrmions

Positionally disordered but orientationally ordered

See also: Bowick & Giomi. Advances in Physics, 2009
Giomi, Toner & Sarkar. Phys Rev E 2022

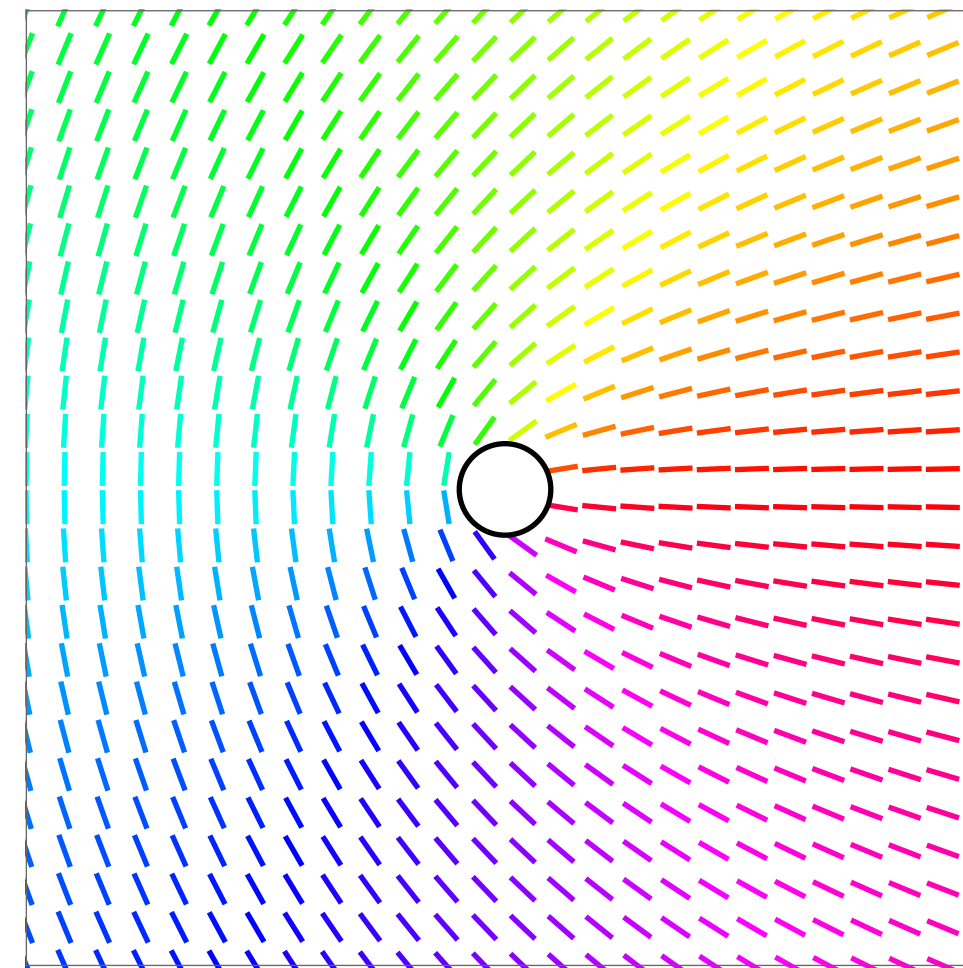
Topological defects in nematic liquid crystals



Uni-axial nematic LC

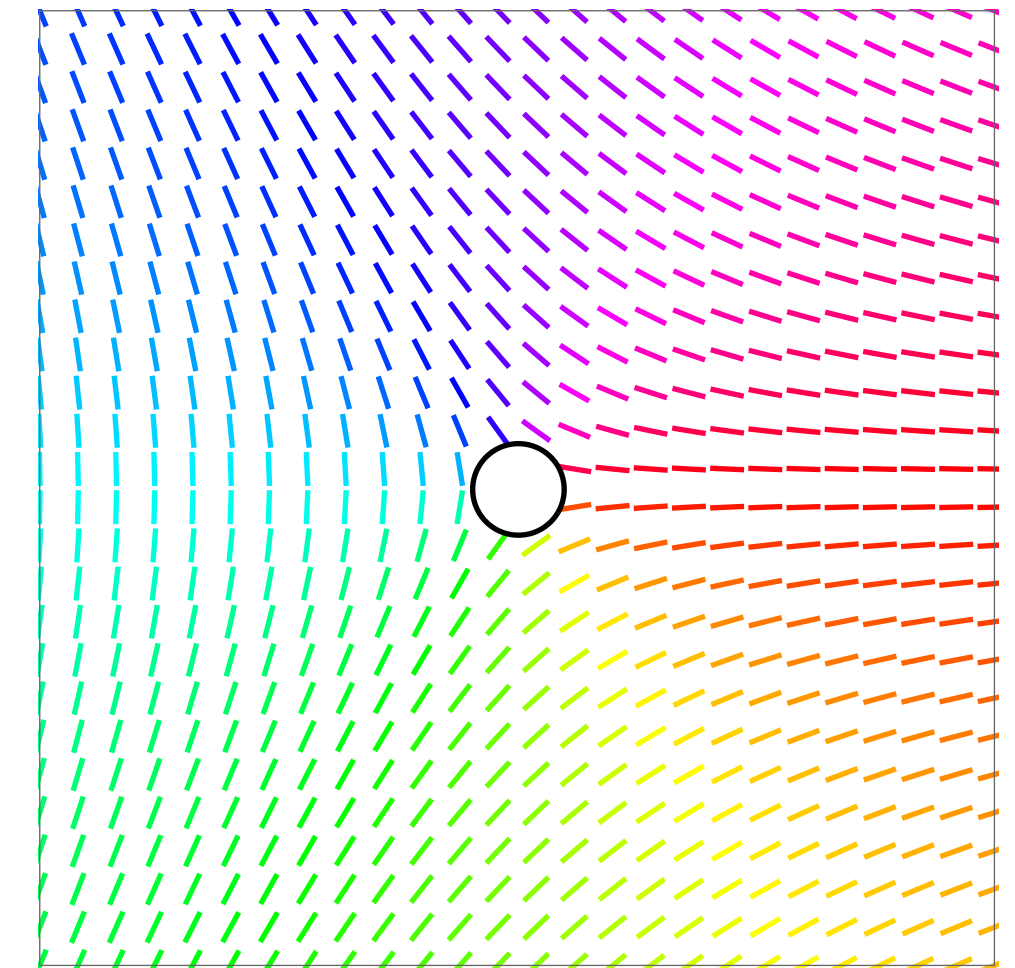
$$\mathbf{Q}(\mathbf{r}) = S(\mathbf{r}) \left[\mathbf{n}(\mathbf{r}) \otimes \mathbf{n}(\mathbf{r}) - \frac{\mathbf{I}}{d} \right]$$

$q = +1/2$



$$\mathbf{n}_+(\mathbf{r}) = \left(\cos \frac{\phi}{2}, \sin \frac{\phi}{2}, 0 \right)$$

$q = -1/2$



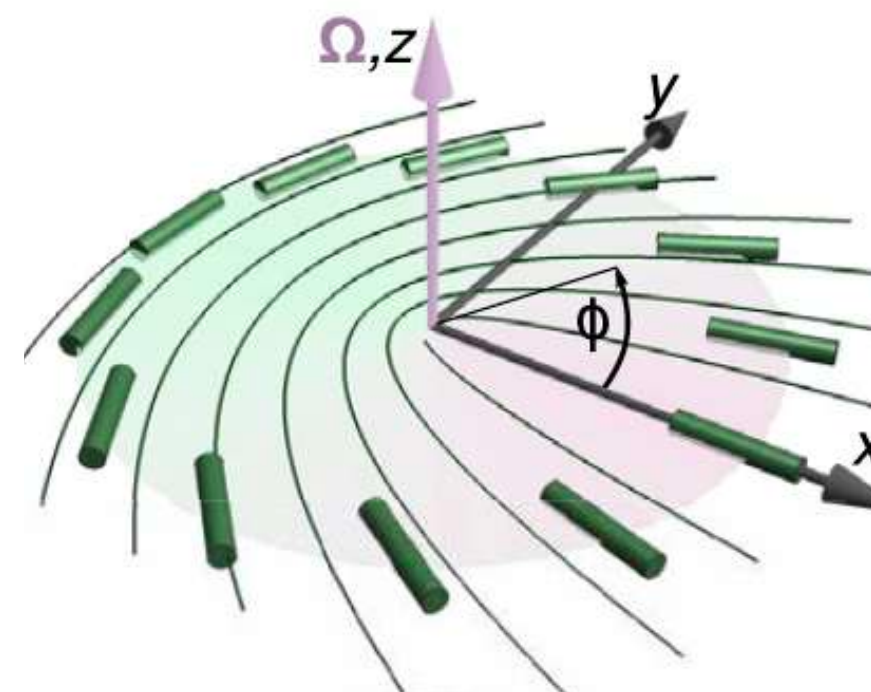
$$\mathbf{n}_-(\mathbf{r}) = \left(\cos \frac{\phi}{2}, \sin \frac{-\phi}{2}, 0 \right)$$

Topological defects in nematic liquid crystals

Uni-axial nematic LC

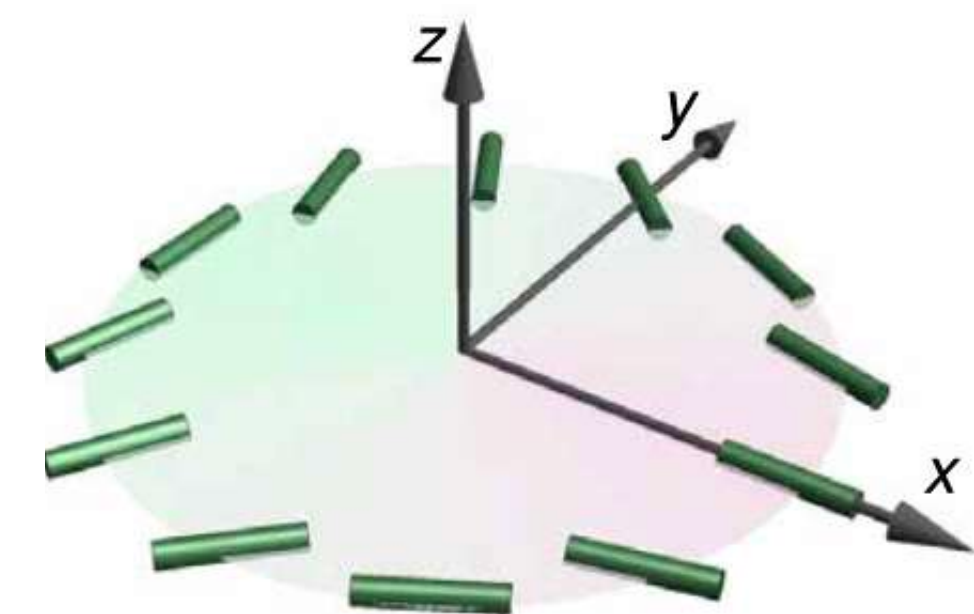
$$\mathbf{Q}(\mathbf{r}) = S(\mathbf{r}) \left[\mathbf{n}(\mathbf{r}) \otimes \mathbf{n}(\mathbf{r}) - \frac{\mathbf{I}}{d} \right]$$

$$q = +1/2$$

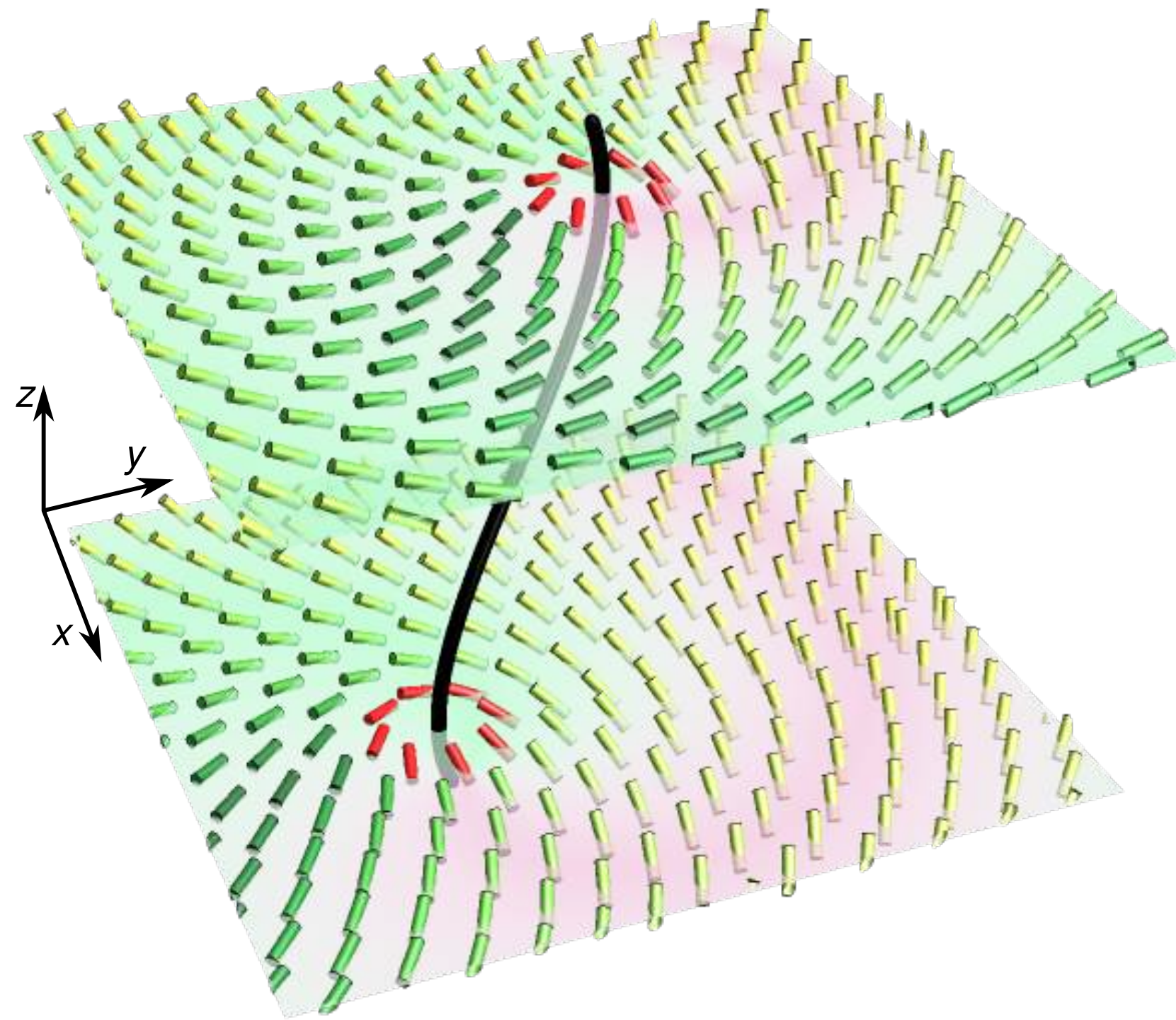


$$\mathbf{n}_+(\mathbf{r}) = \left(\cos \frac{\phi}{2}, \sin \frac{\phi}{2}, 0 \right)$$

$$q = -1/2$$



$$\mathbf{n}_-(\mathbf{r}) = \left(\cos \frac{\phi}{2}, \sin \frac{-\phi}{2}, 0 \right)$$



Nematic bits (nbits)



Ziga Kos

Write director in Pauli-matrix basis

$$\mathbf{n}_0(r) = \left(\cos \frac{\phi}{2} \right) \boldsymbol{\sigma}_x + \left(\sin \frac{\phi}{2} \right) \boldsymbol{\sigma}_y + 0 \boldsymbol{\sigma}_z$$



Other director fields obtained by quaternion transformation

$$\mathbf{n}(\mathbf{r}) = \eta \mathbf{n}_0(\mathbf{r}) \eta^\dagger$$

Unit-norm quaternion

$$\eta = \left(\cos \frac{\theta}{2} \right) \mathbf{1} + i \left(\sin \frac{\theta}{2} \right) (\hat{a}_x \boldsymbol{\sigma}_x + \hat{a}_y \boldsymbol{\sigma}_y + \hat{a}_z \boldsymbol{\sigma}_z)$$

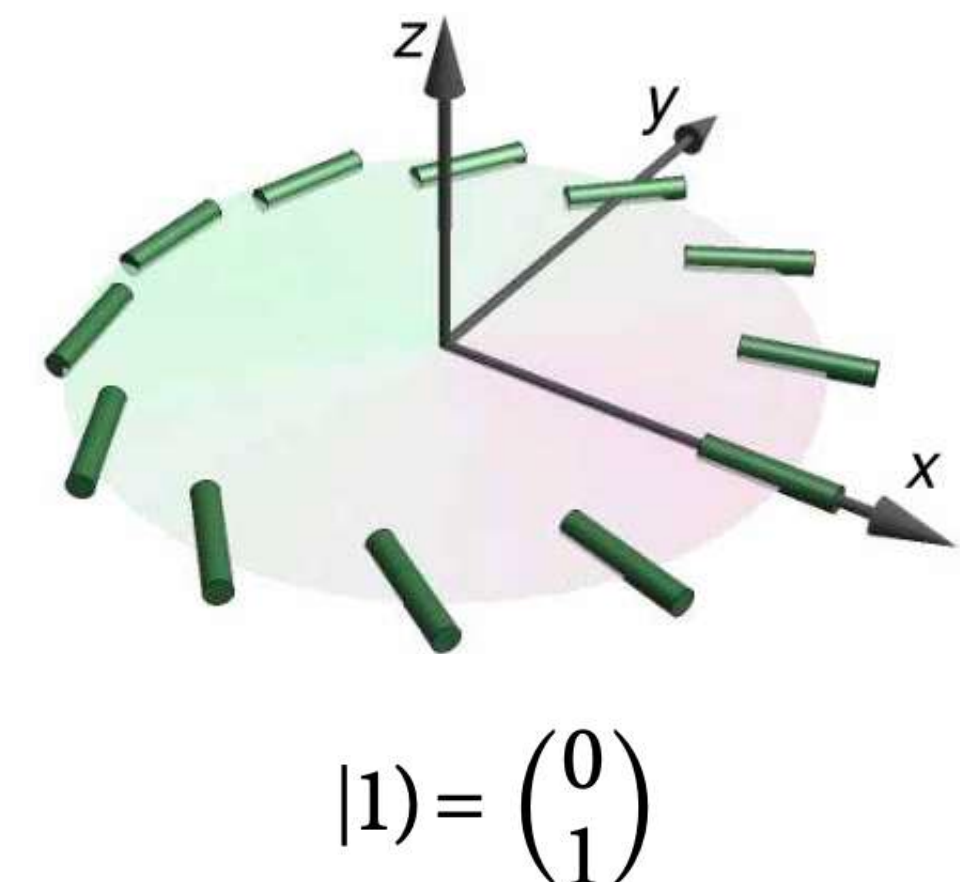
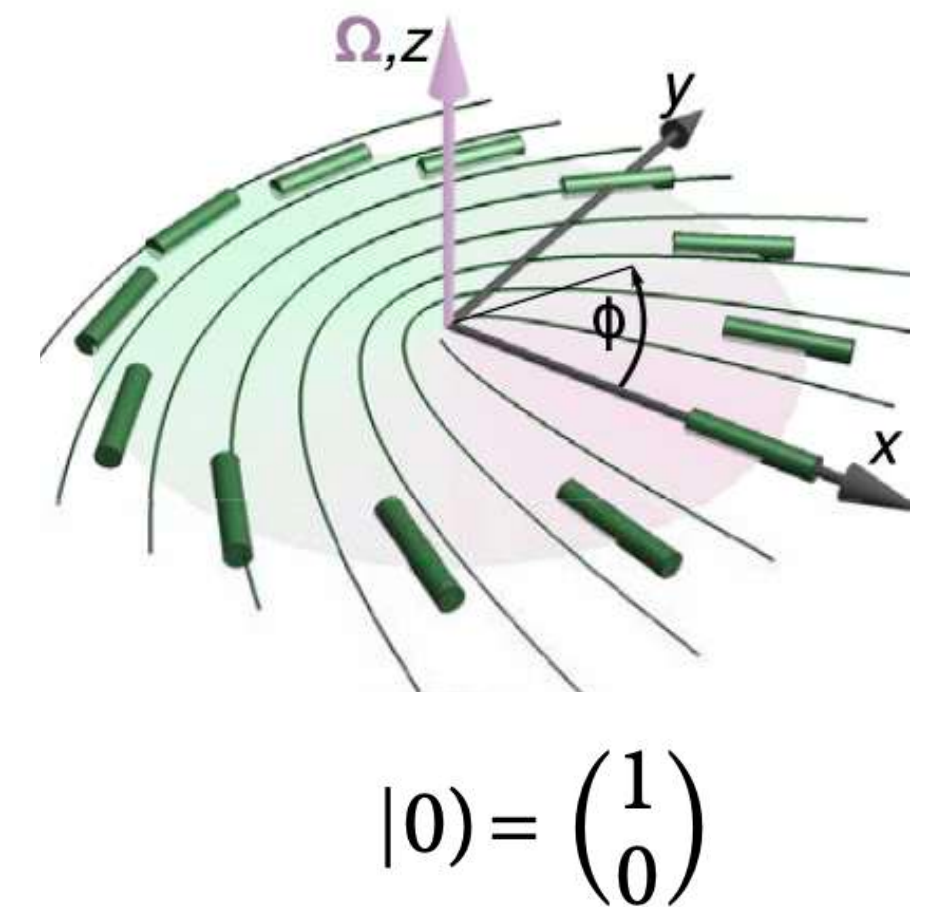


Unit-norm quaternions equivalent to SU(2)

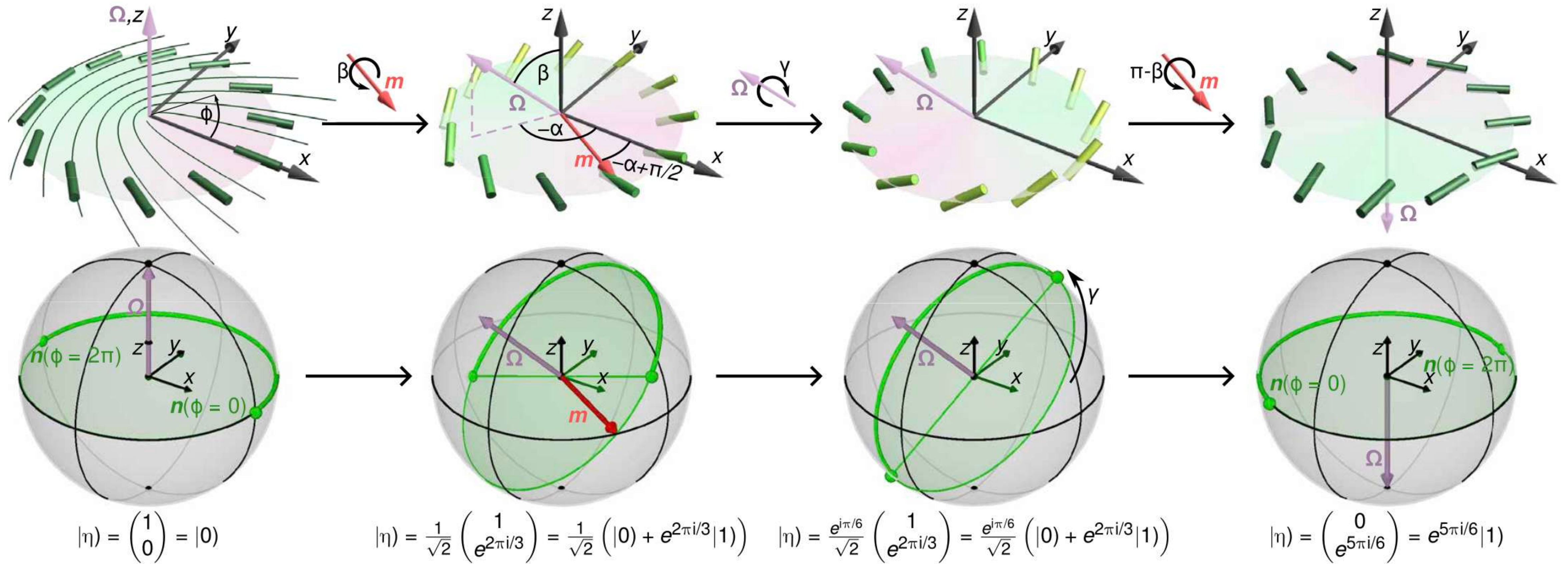
$$\begin{pmatrix} c_1 & -c_2^* \\ c_2 & c_1^* \end{pmatrix} \quad |c_1|^2 + |c_2|^2 = 1$$

Fully specified by first column

$$|\eta\rangle = c_1 |0\rangle + c_2 |1\rangle = e^{i\frac{\gamma}{2}} \left[\cos \frac{\beta}{2} |0\rangle + e^{-i\alpha} \sin \frac{\beta}{2} |1\rangle \right]$$



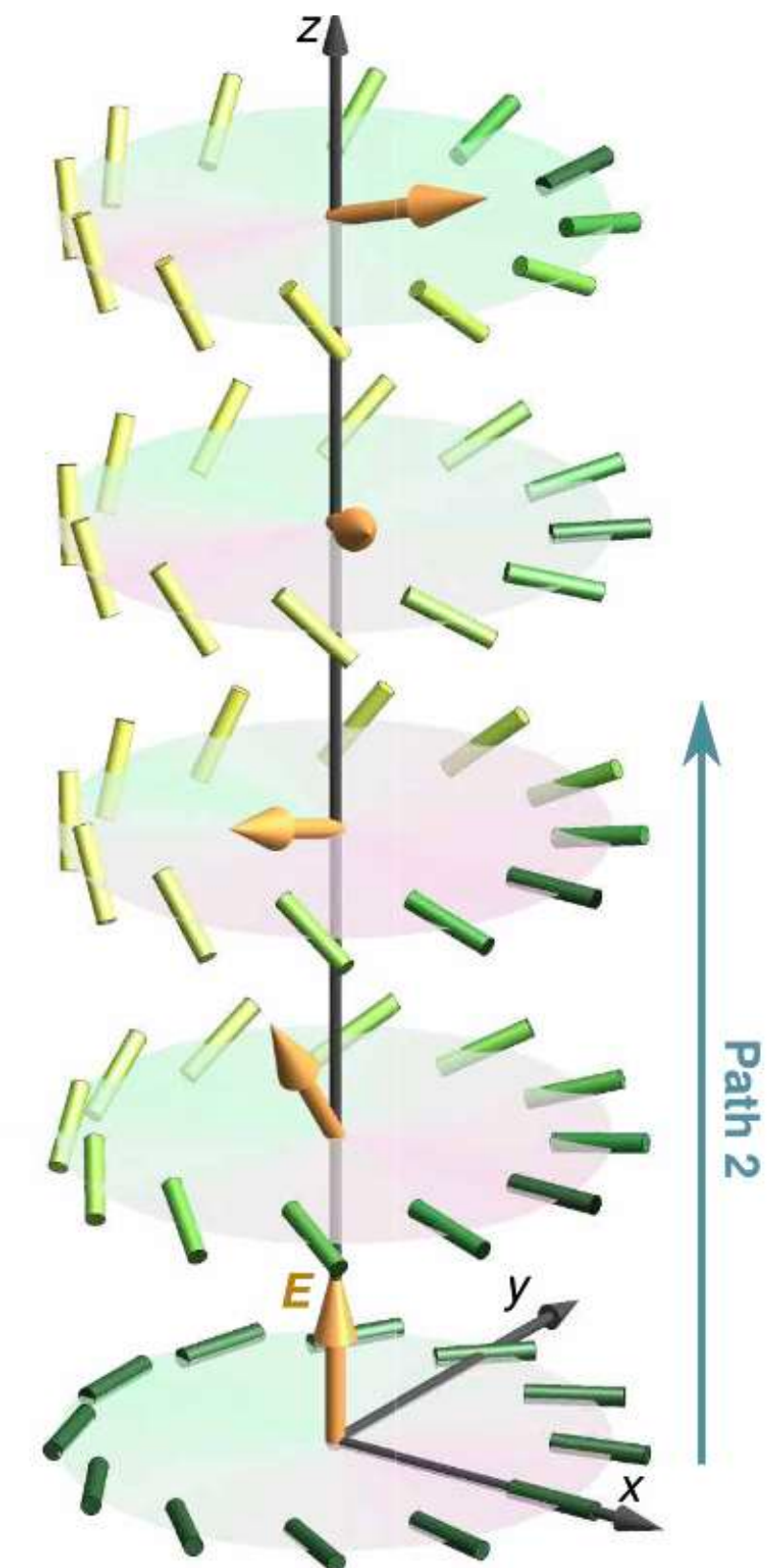
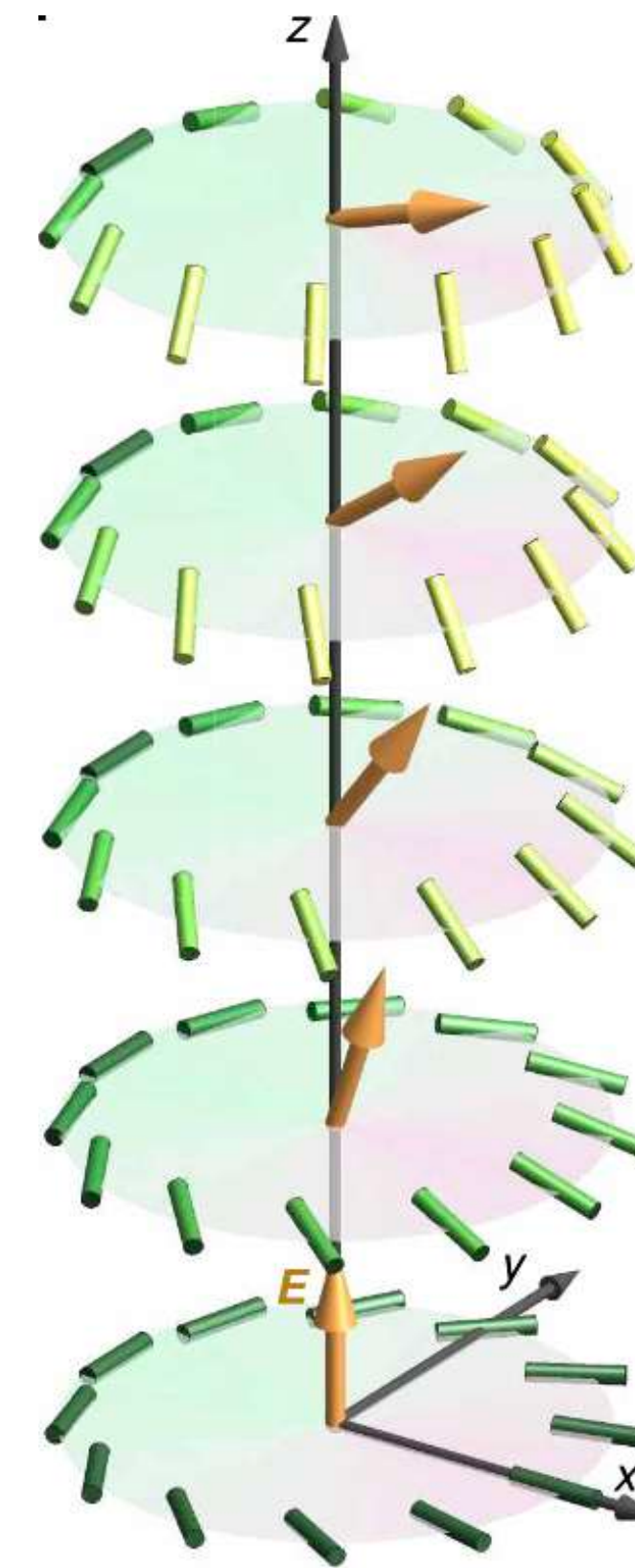
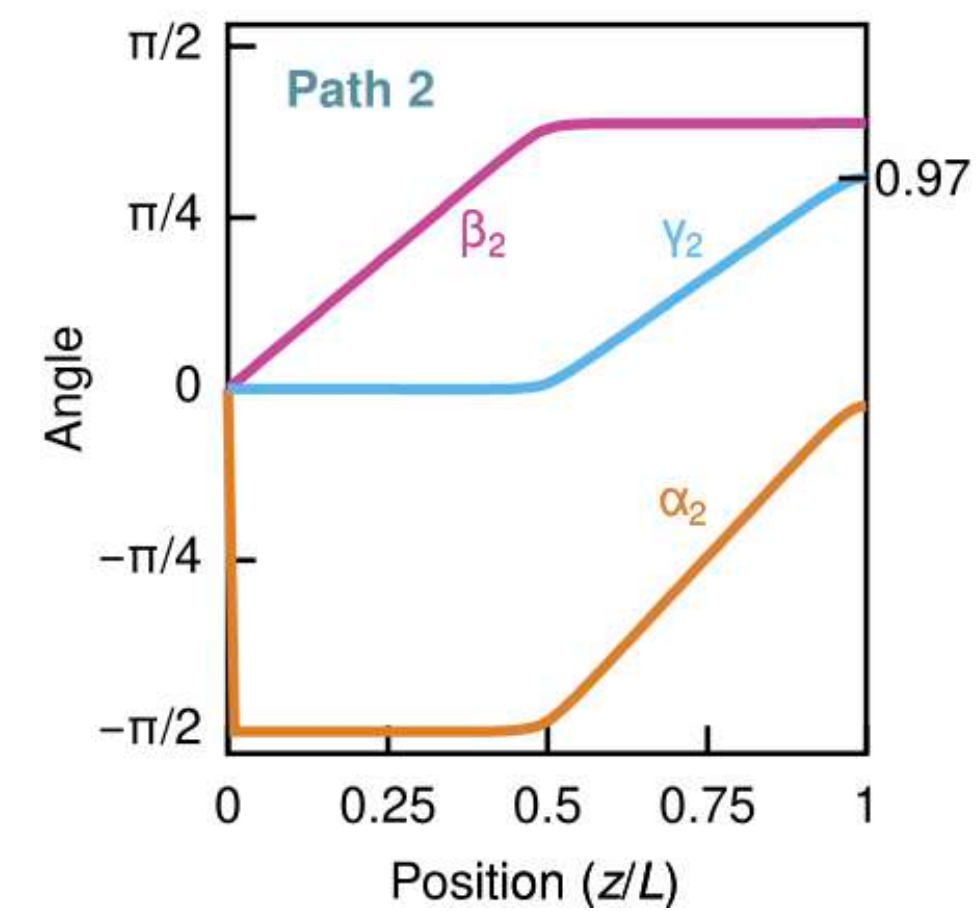
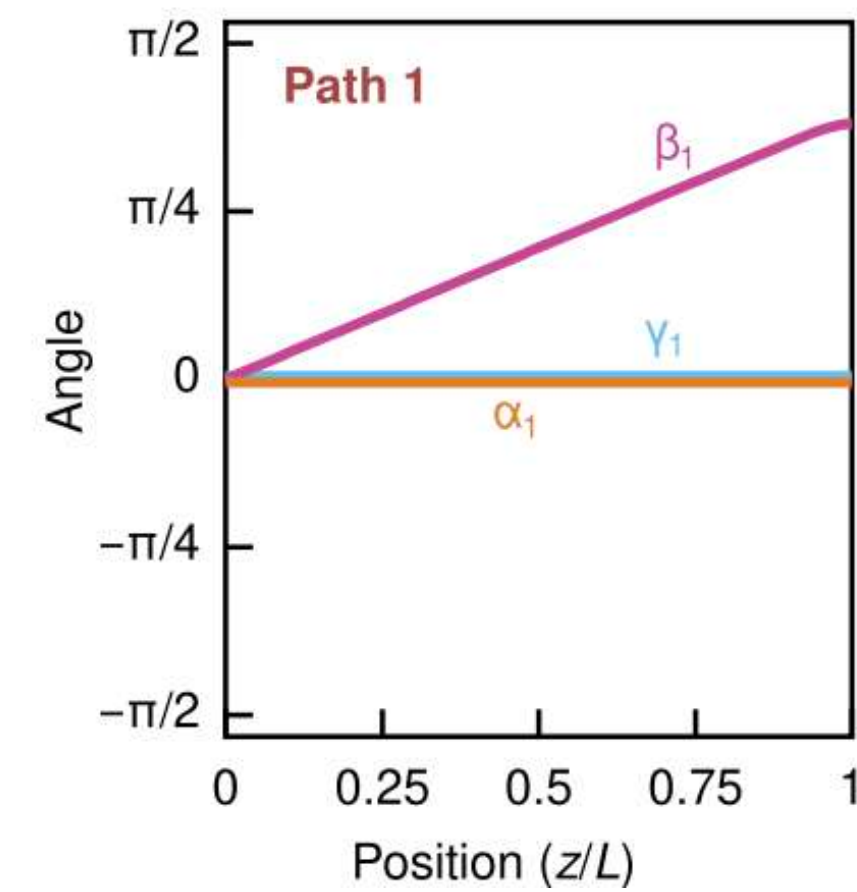
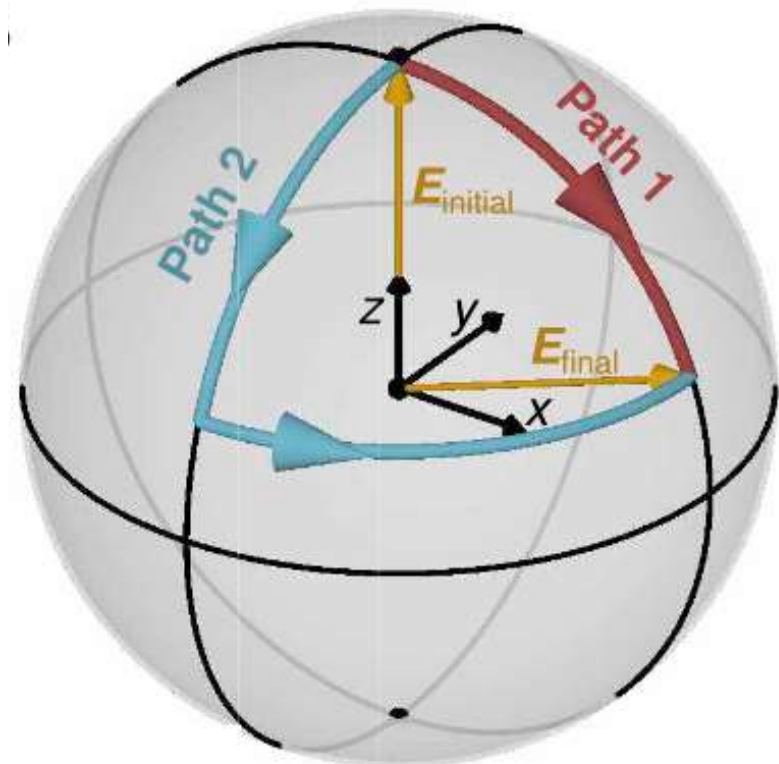
$$|\eta\rangle = c_1 |0\rangle + c_2 |1\rangle = e^{i\frac{\gamma}{2}} \left[\cos \frac{\beta}{2} |0\rangle + e^{-i\alpha} \sin \frac{\beta}{2} |1\rangle \right]$$



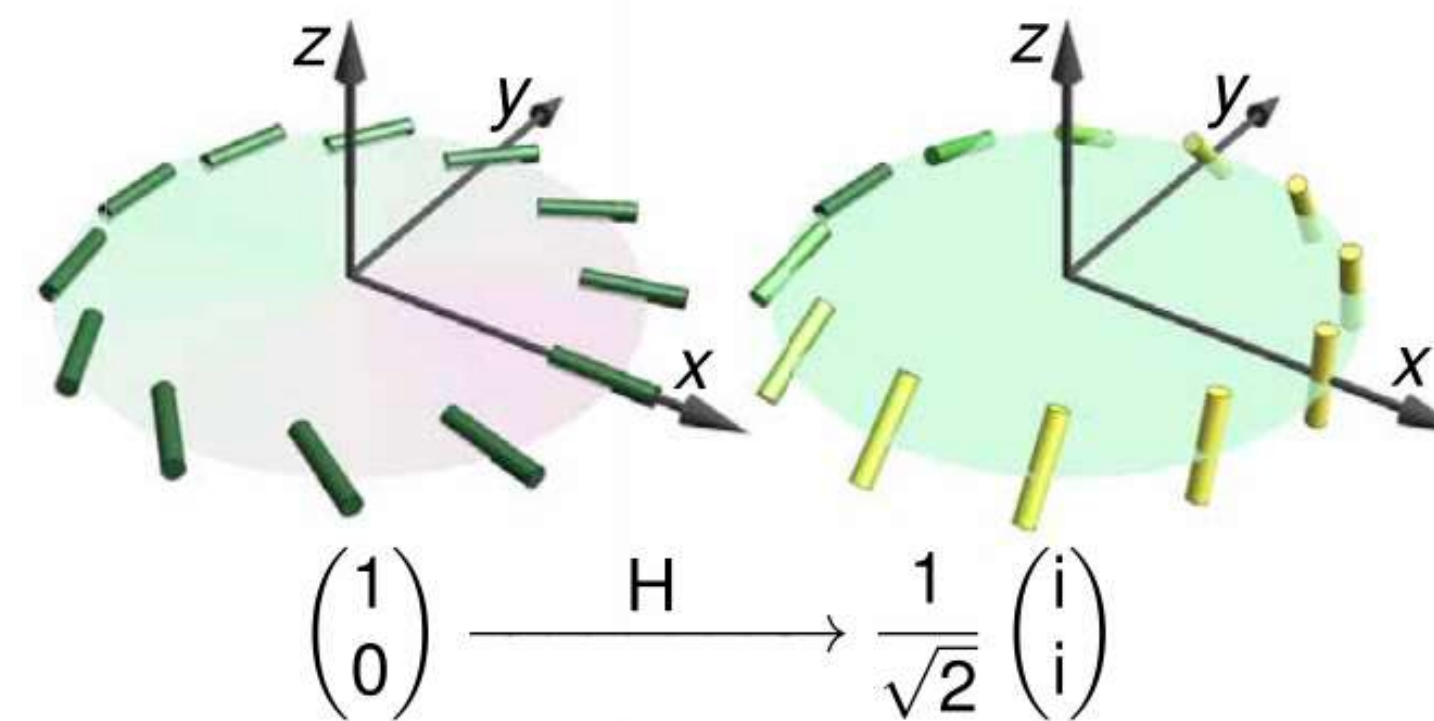
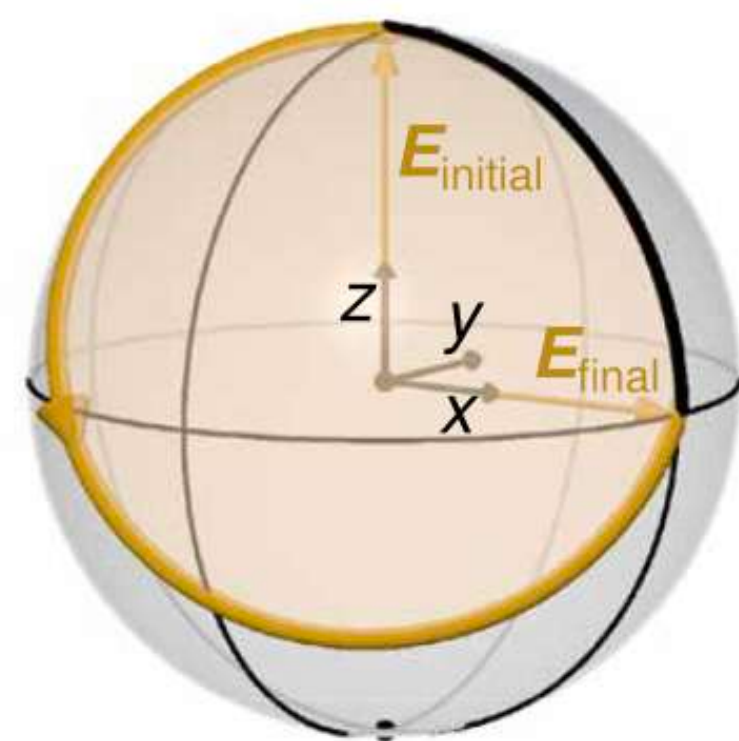
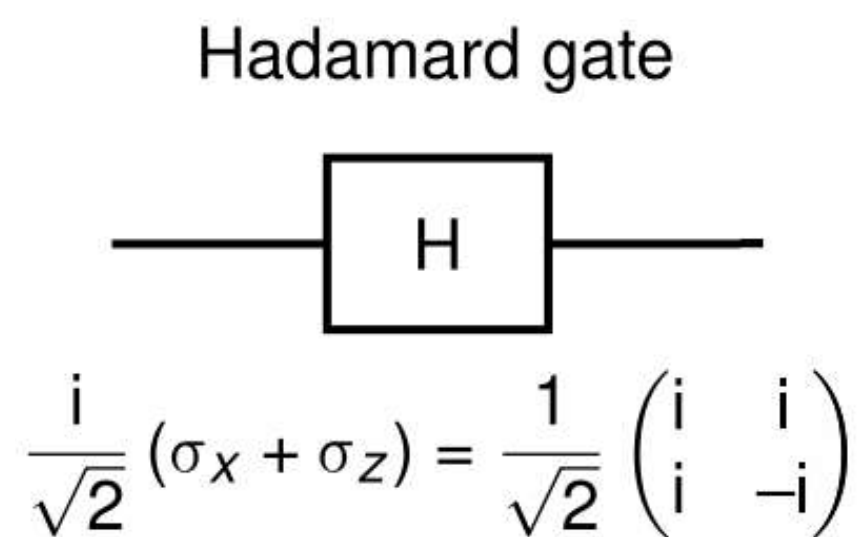
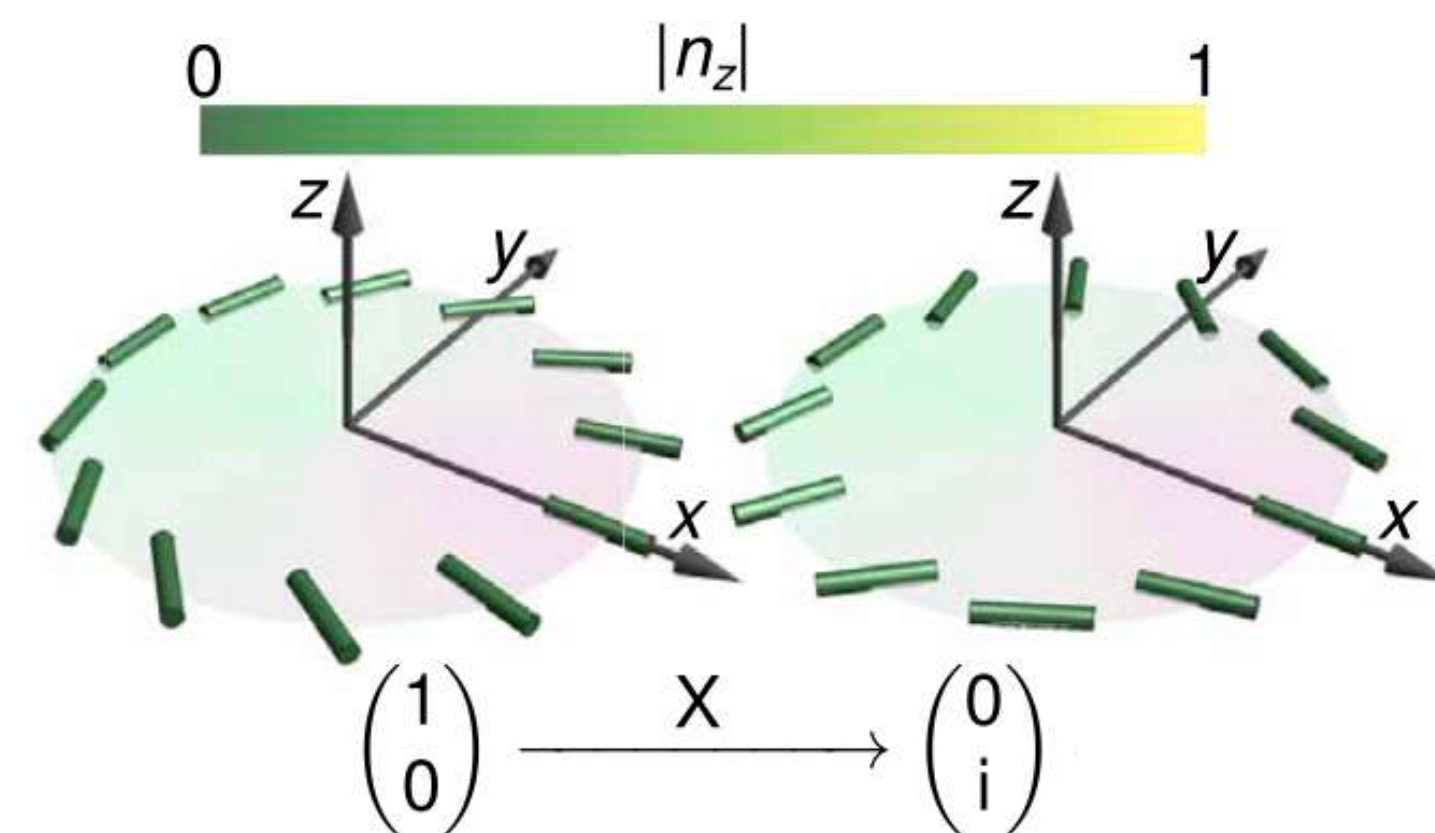
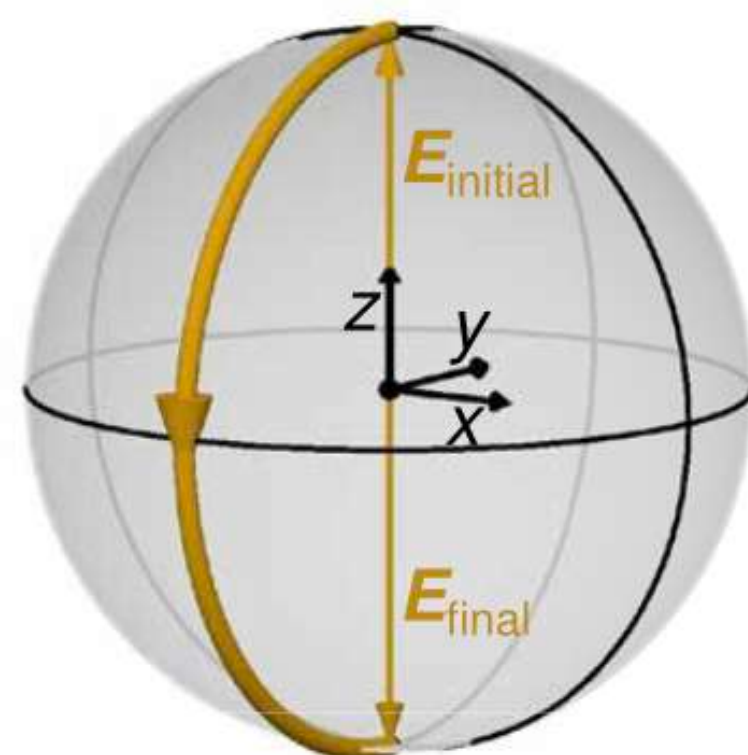
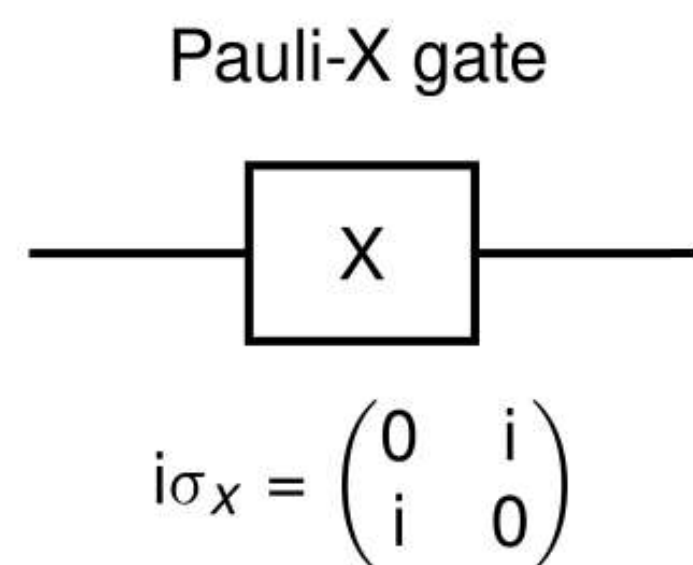
nbit manipulations with electric fields



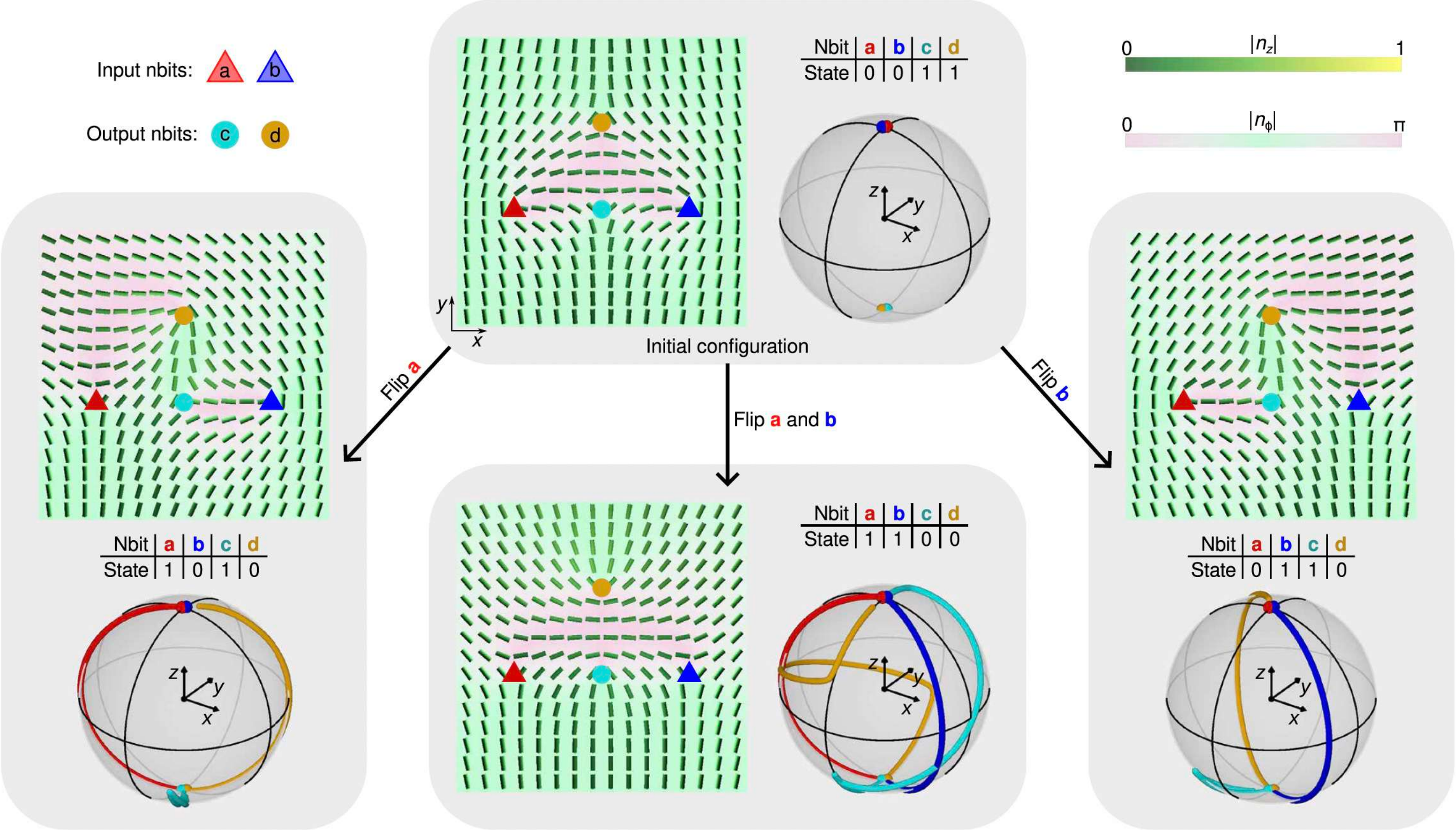
$$\frac{d\eta}{dt} = \frac{K}{\Gamma} \frac{\partial^2 \eta}{\partial z^2} - \frac{\epsilon_a}{4\Gamma} \text{Tr}(\mathbf{E}\eta\sigma_z\eta^\dagger) \mathbf{E}\eta\sigma_z - \tilde{\lambda}\eta, \quad \tilde{\lambda} = \frac{\lambda}{\pi\Gamma}$$



single nbit-operations



Classical nbit-logic : universal gates



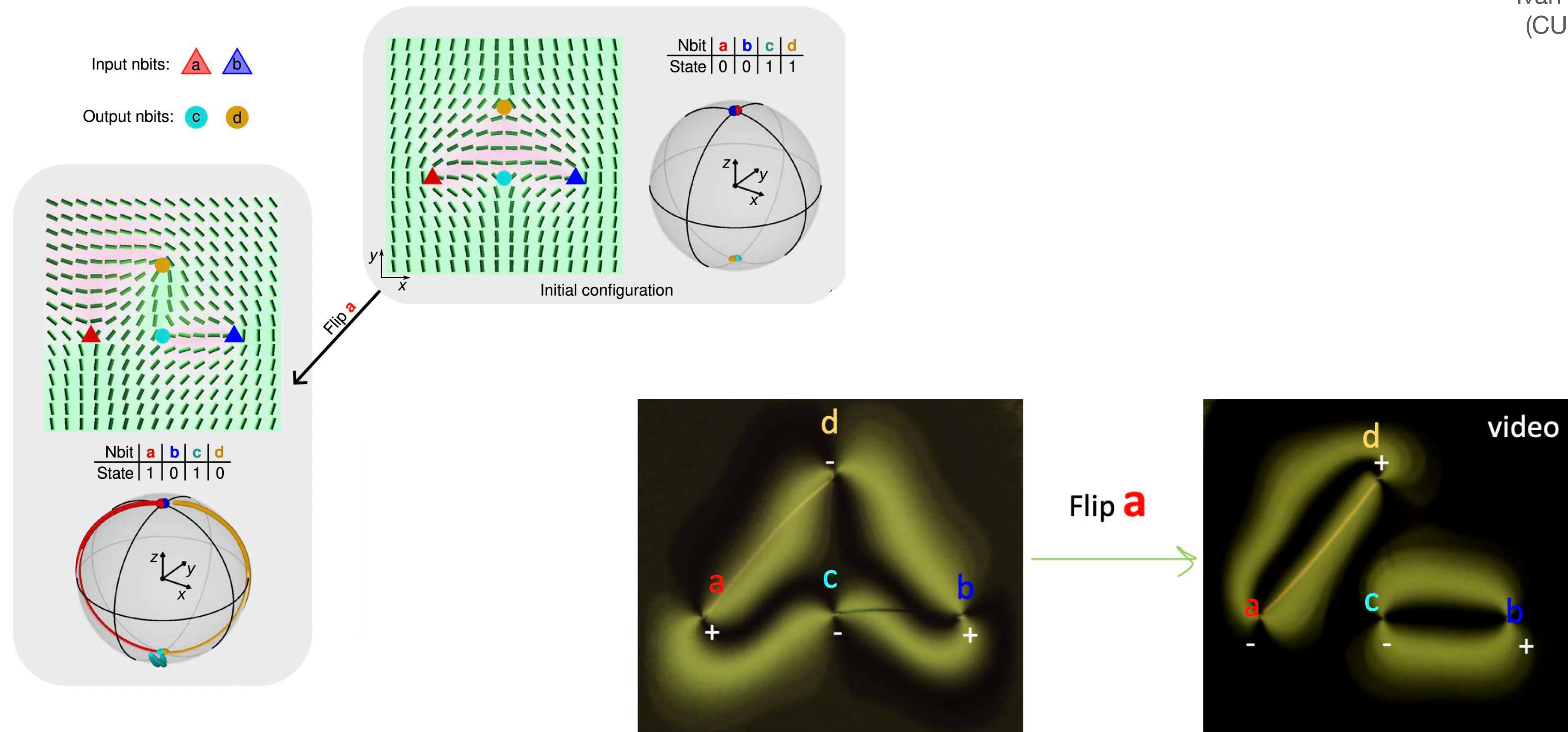
c: NAND

d: NOR

Recent experiments: towards nbit logic gates



Ivan Smalyukh
(CU Boulder)



Recent experiments: towards nbit memory



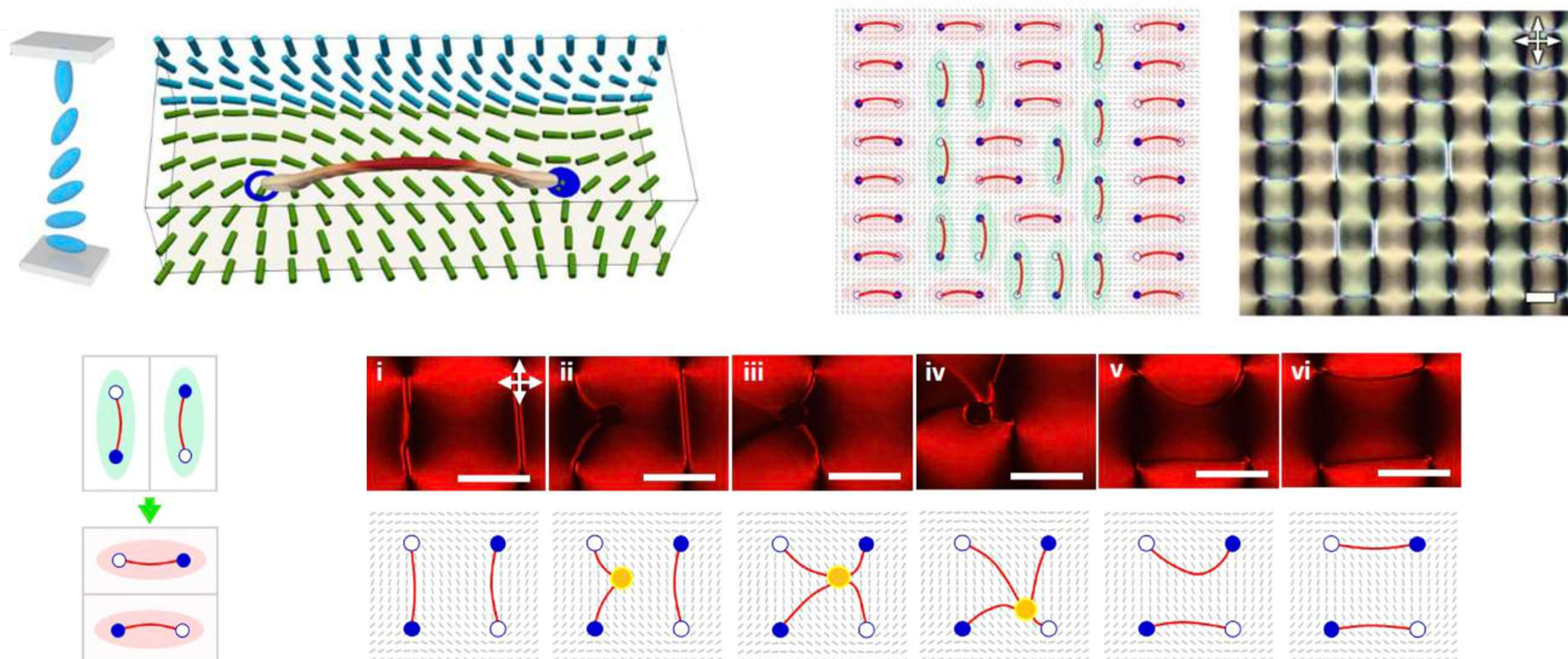
Ivan Smalyukh
(CU Boulder)

PHYSICAL REVIEW X **15**, 021084 (2025)

Featured in Physics

Emergent Dimer-Model Topological Order and Quasiparticle Excitations in Liquid Crystals: Combinatorial Vortex Lattices

Cuiling Meng,¹ Jin-Sheng Wu¹, Žiga Kos^{2,3,4,5}, Jörn Dunkel^{2,3}, Cristiano Nisoli^{6,*} and Ivan I. Smalyukh^{1,2,7,8,†}

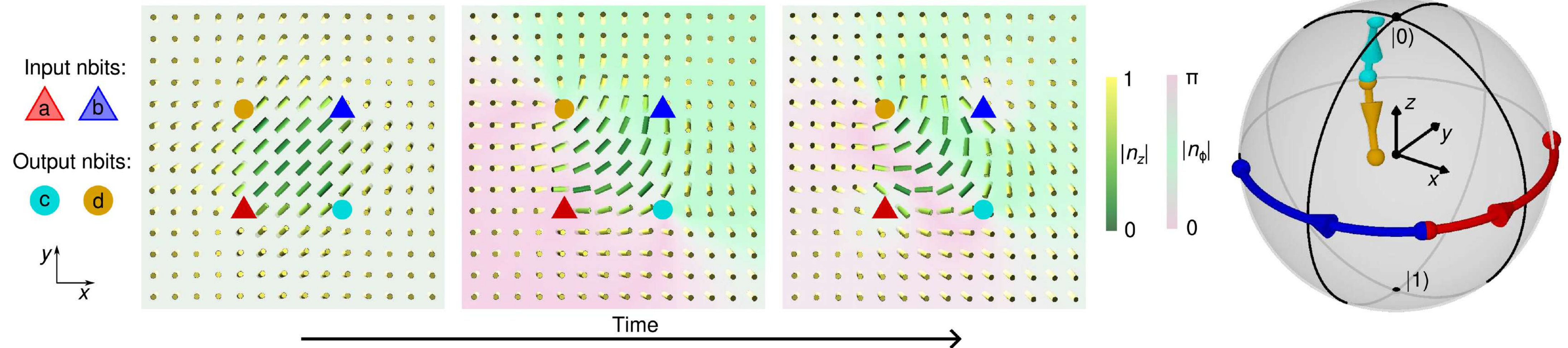


Outlook: beyond classical bit logic



Ziga Kos

$$|\eta\rangle = c_1 |0\rangle + c_2 |1\rangle = e^{i\frac{\gamma}{2}} \left[\cos \frac{\beta}{2} |0\rangle + e^{-i\alpha} \sin \frac{\beta}{2} |1\rangle \right]$$



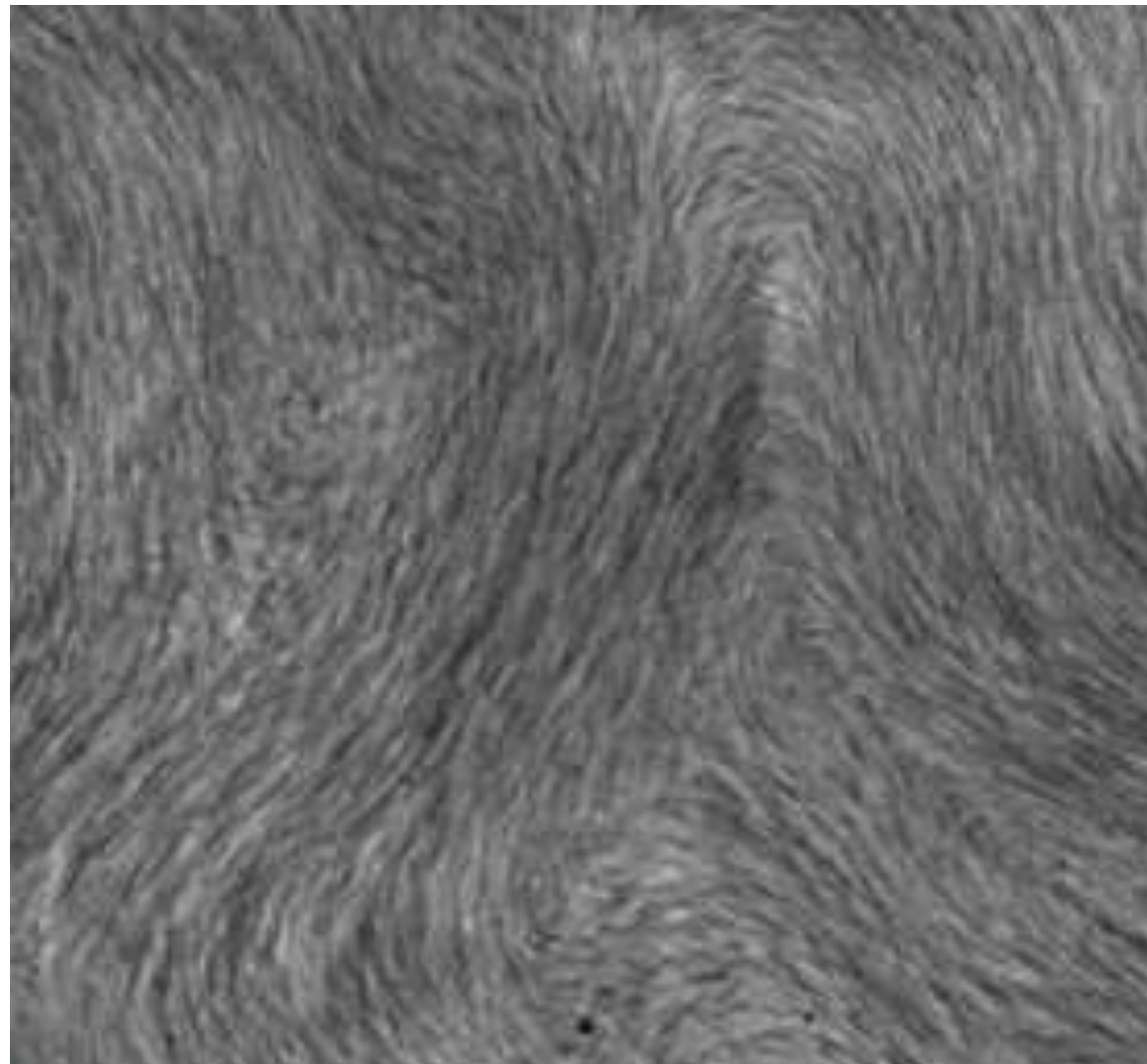
What is the space of controllable non-classical bit gate operations ?



Petur Bryde

Two-dimensional dihedral ('k-atic') liquid crystals

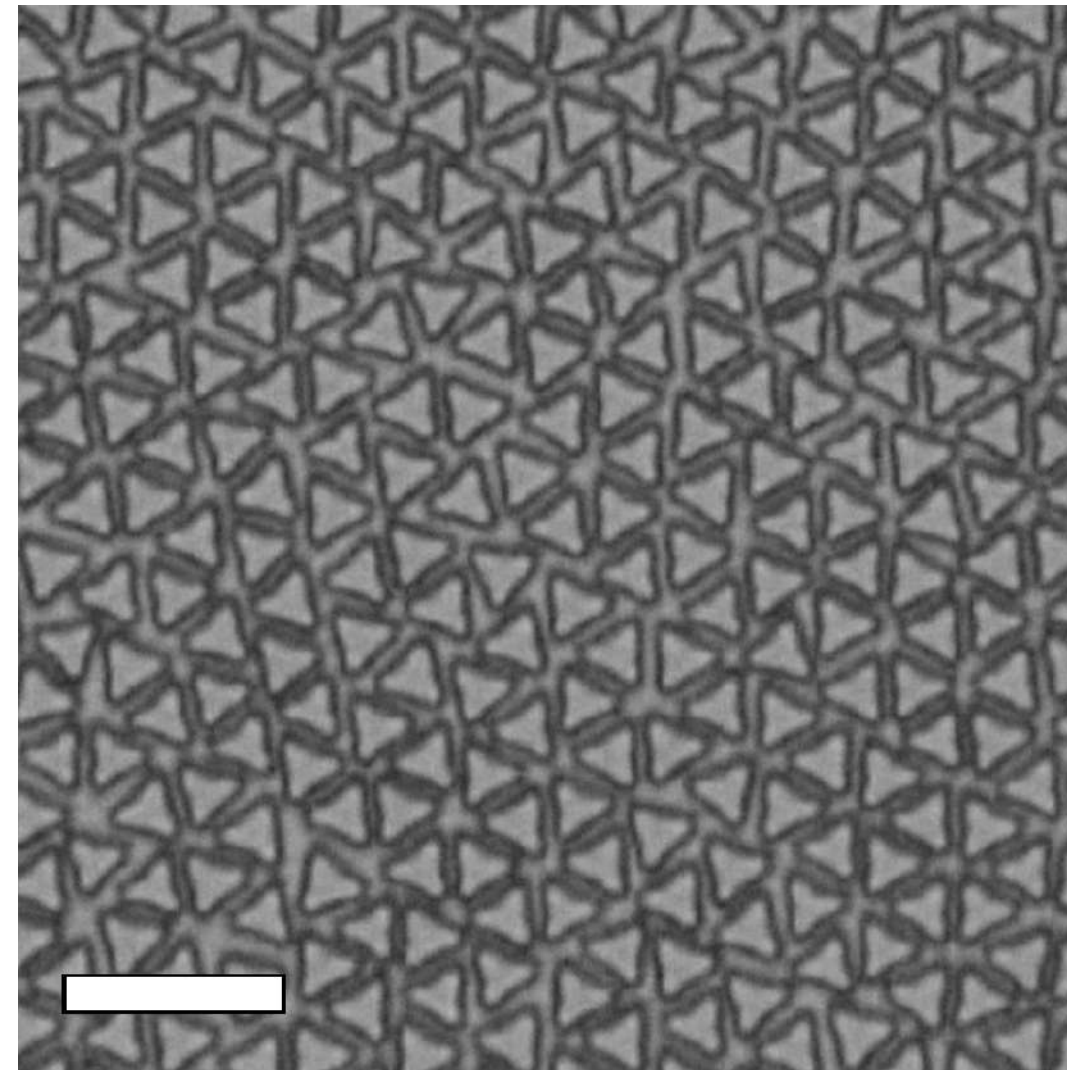
2-atic
(nematic)



Zhou et al. PNAS 2014

bacteria

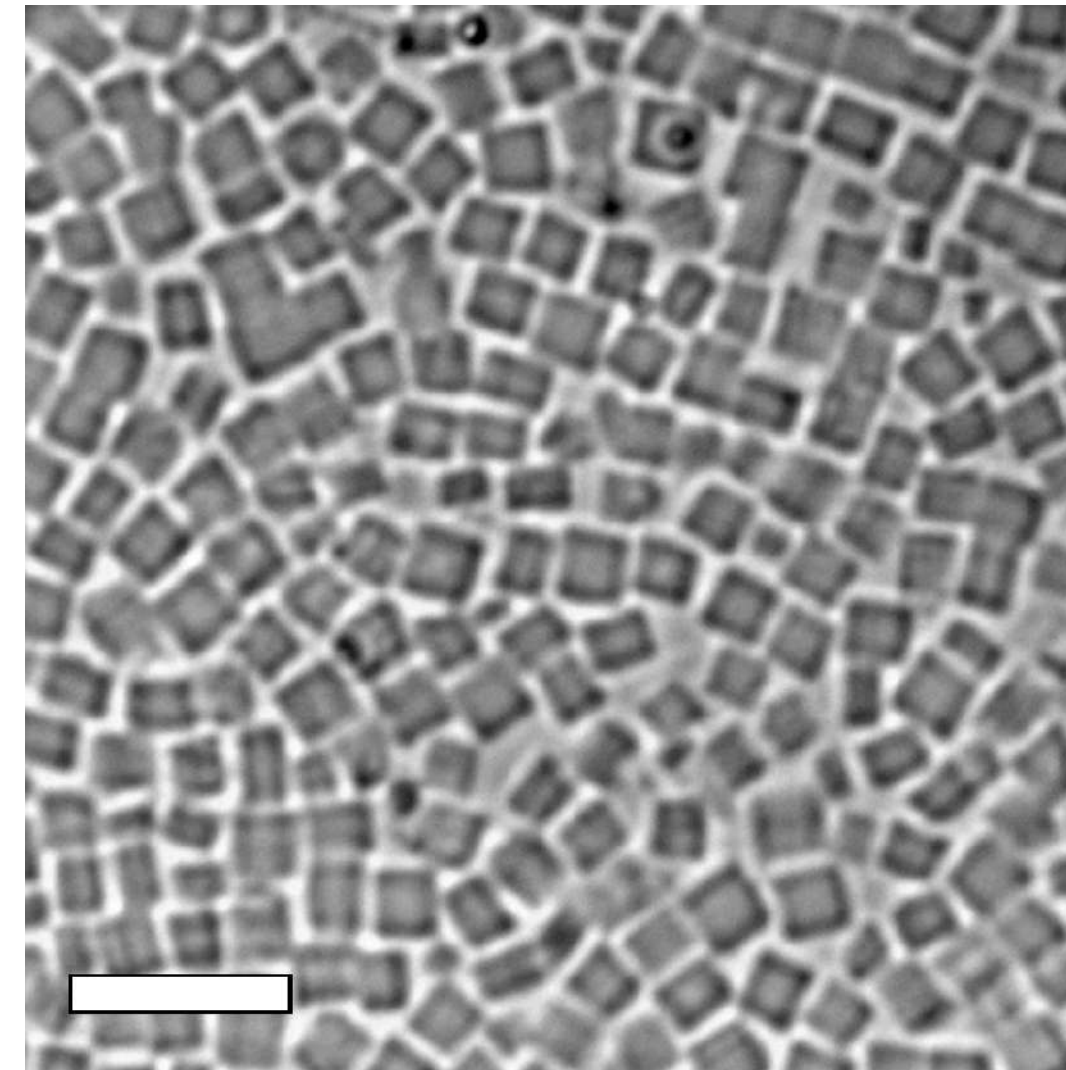
3-atic



Zhao et al. Nat Comm 2012

colloids

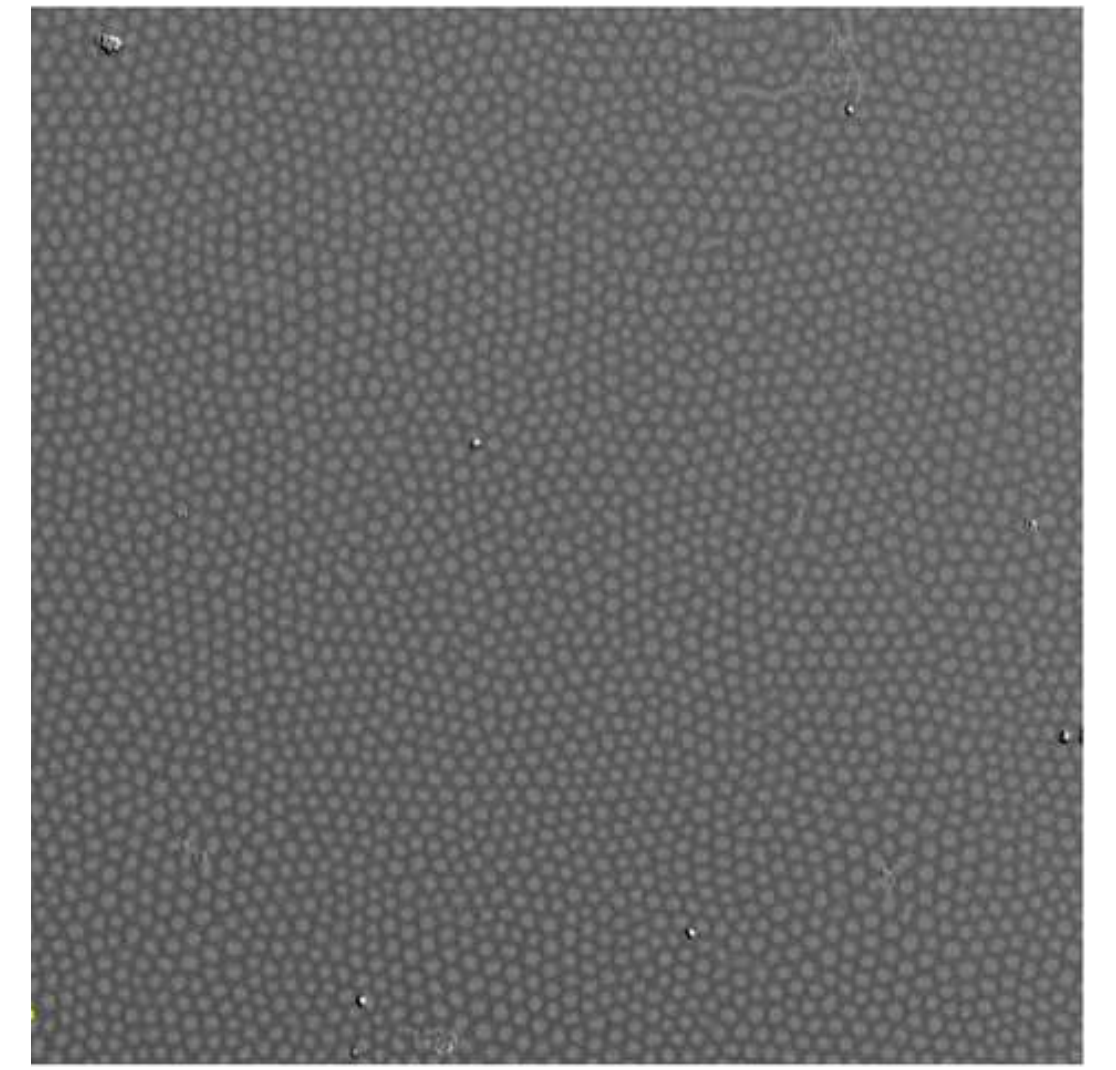
4-atic



Löffler. Thesis 2018 (Konstanz)

colloids

6-atic



Zazvorka et al. Adv Funct Mater 2020

skyrmions

See also: Bowick & Giomi. Advances in Physics, 2009
Giomi, Toner & Sarkar. Phys Rev E 2022

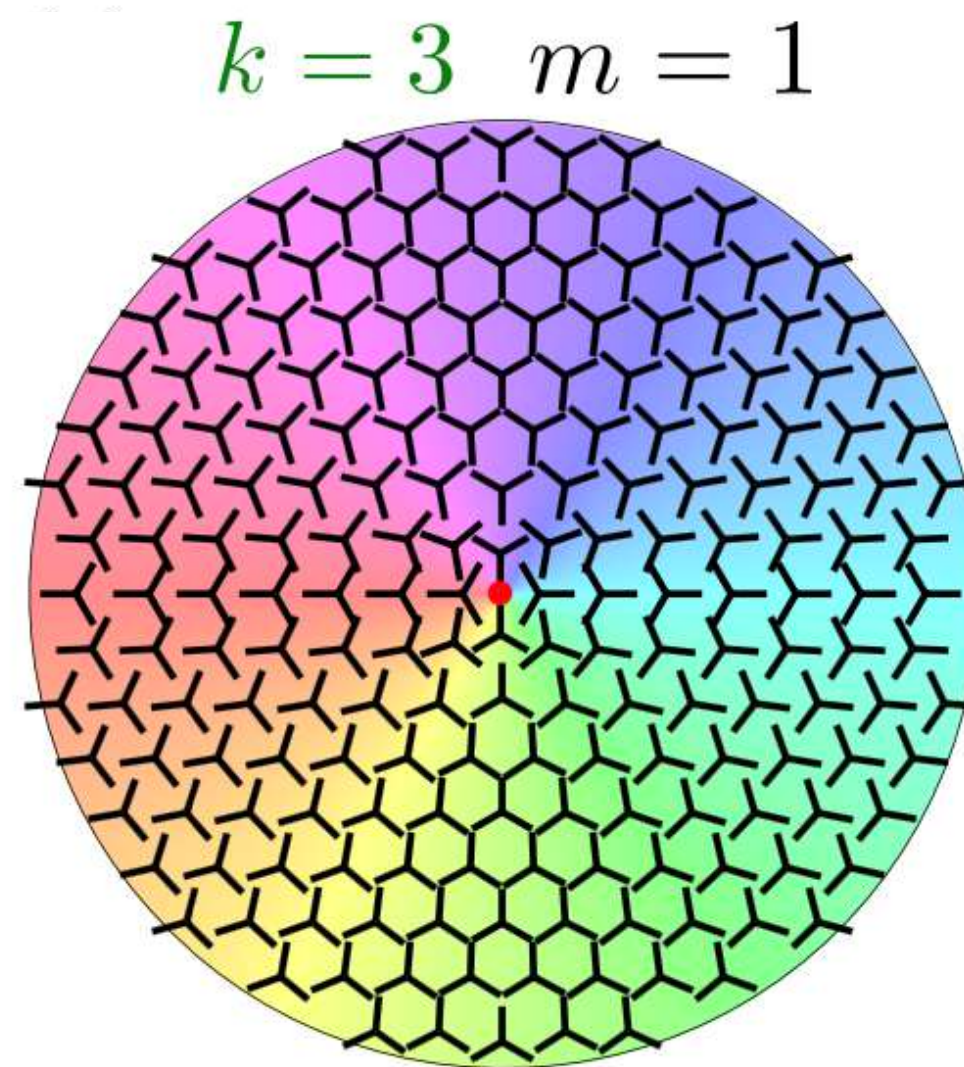
Fractional defects in k-atic LCs

Complex k -atic order parameter

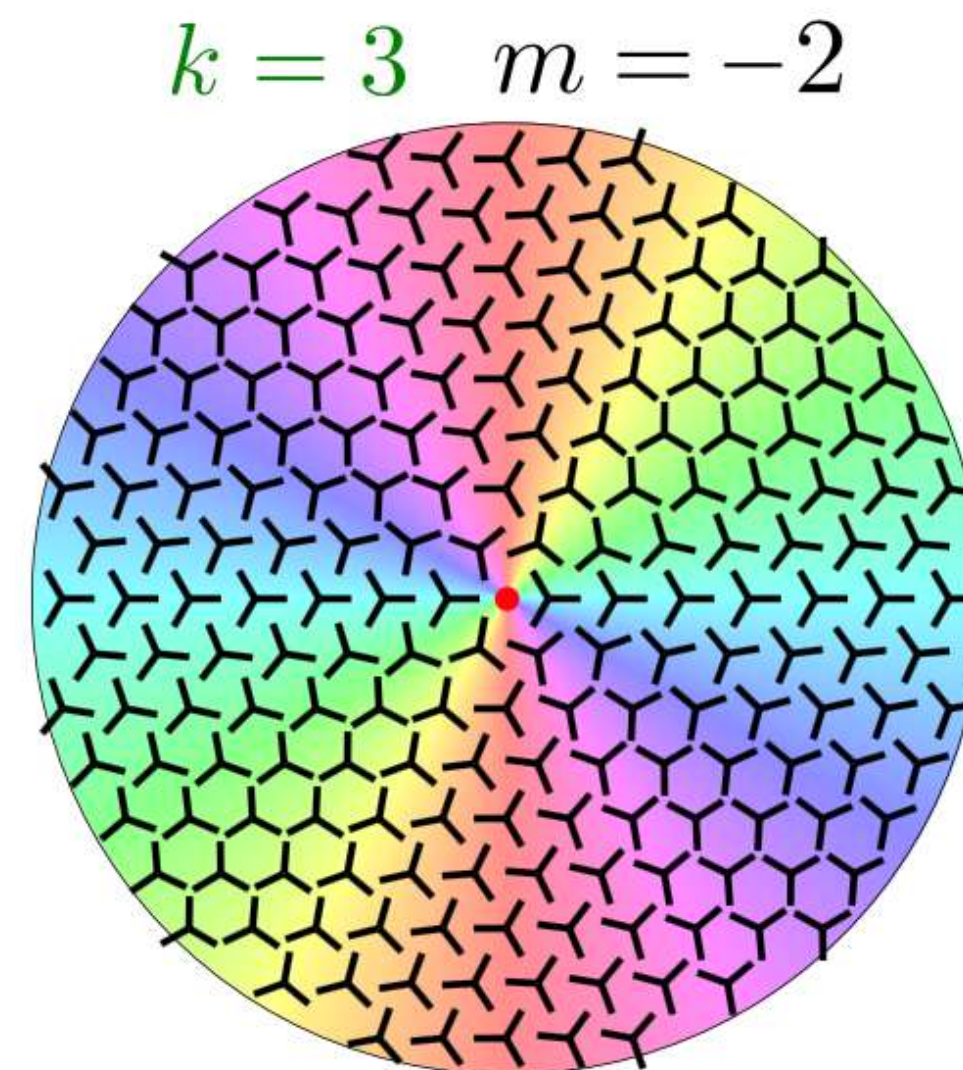
$$\Psi_k = |\Psi_k| e^{ik\phi(k)} \quad k\text{-atic orientation}$$

Topological charge

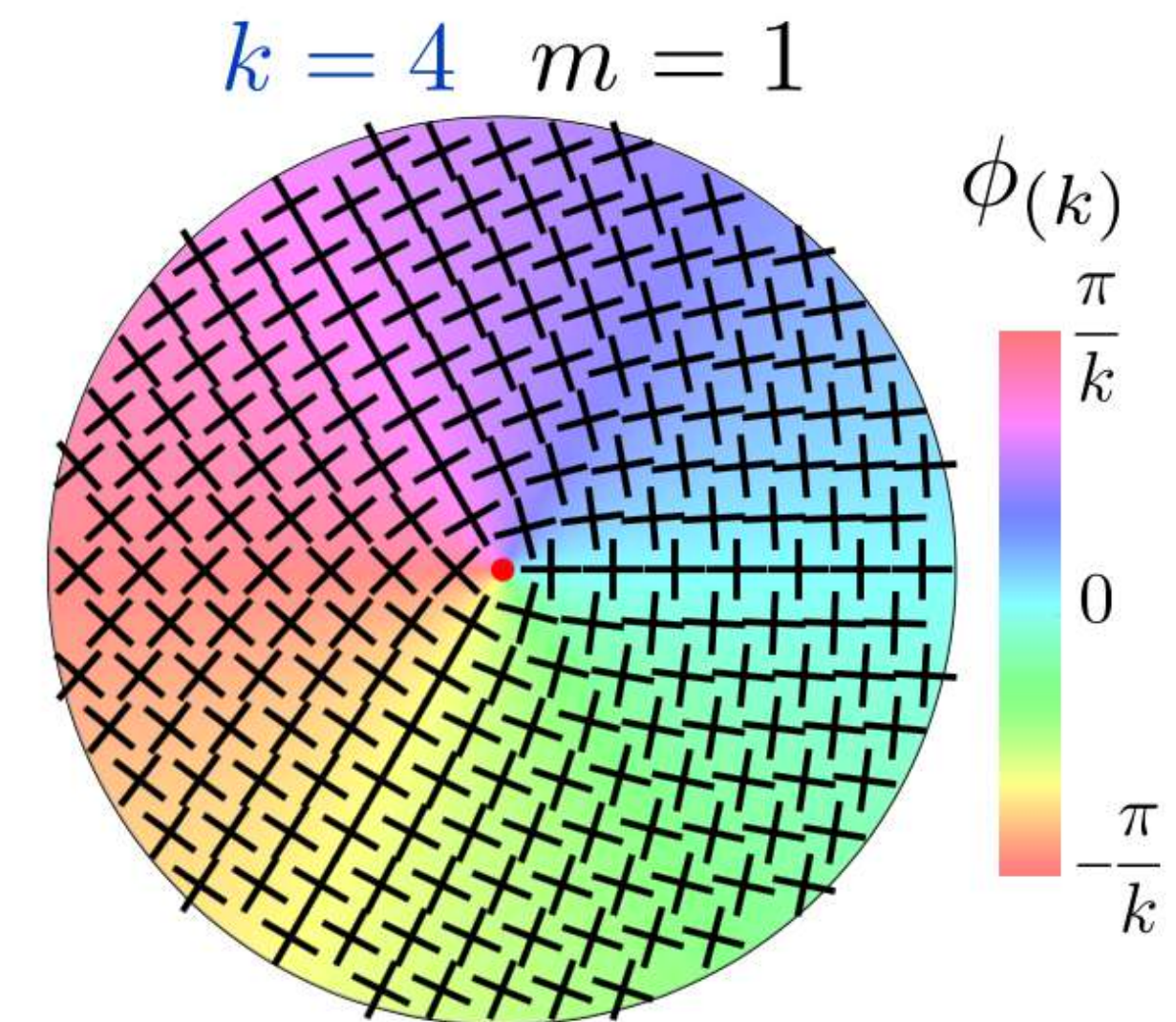
$$q = \frac{1}{2\pi} \oint_{\mathcal{C}} d\mathbf{l} \cdot \nabla \phi(k) = \frac{m}{k}$$



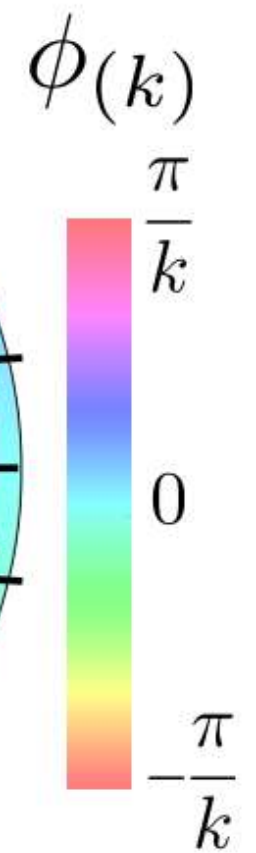
$$q = +1/3$$



$$q = -2/3$$



$$q = +1/4$$



Minimal microscopic disordered-lattice model



Alex Mietke

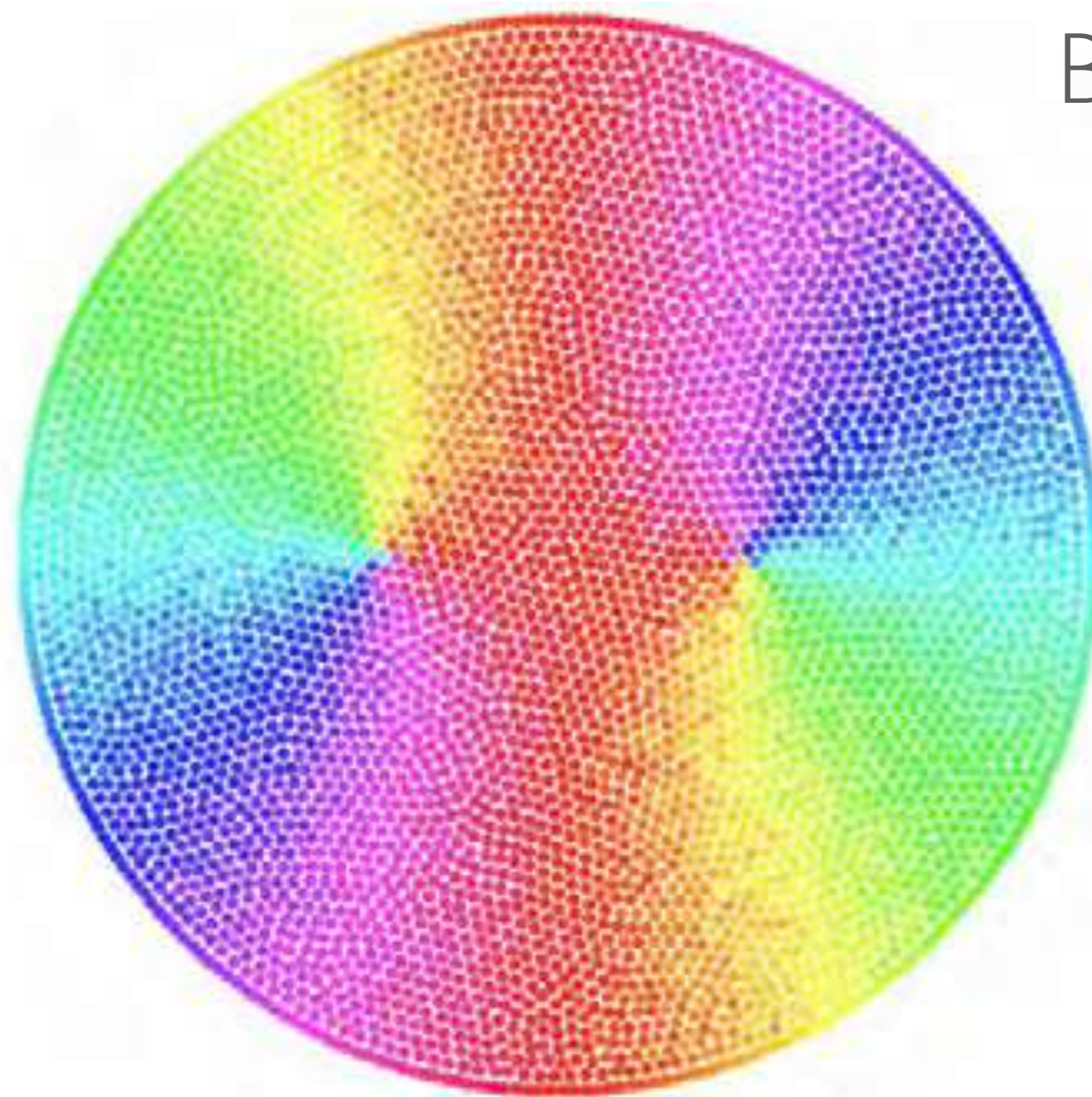
$$\frac{d\alpha_i}{dt} = \frac{g}{\pi R_\alpha^2} \sum_{j \in \mathcal{N}_i} \sin [k(\alpha_j - \alpha_i)] + \sqrt{2D_r} \xi_i(t)$$

$g > 0$: alignment

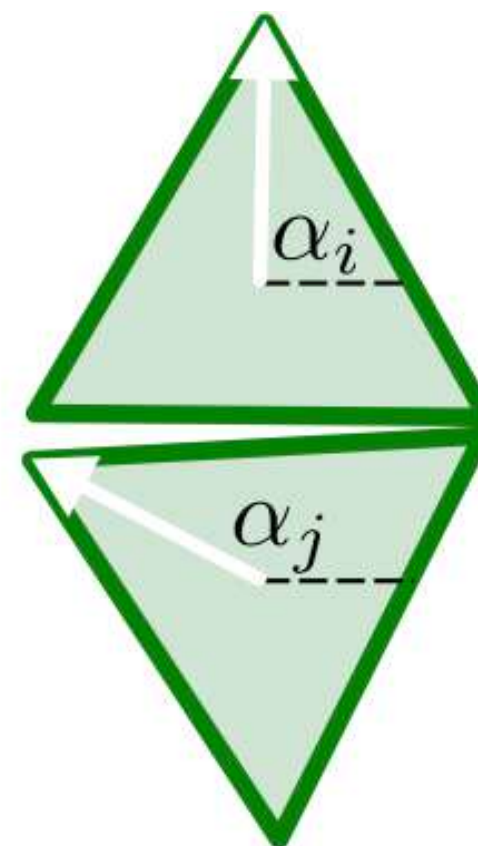
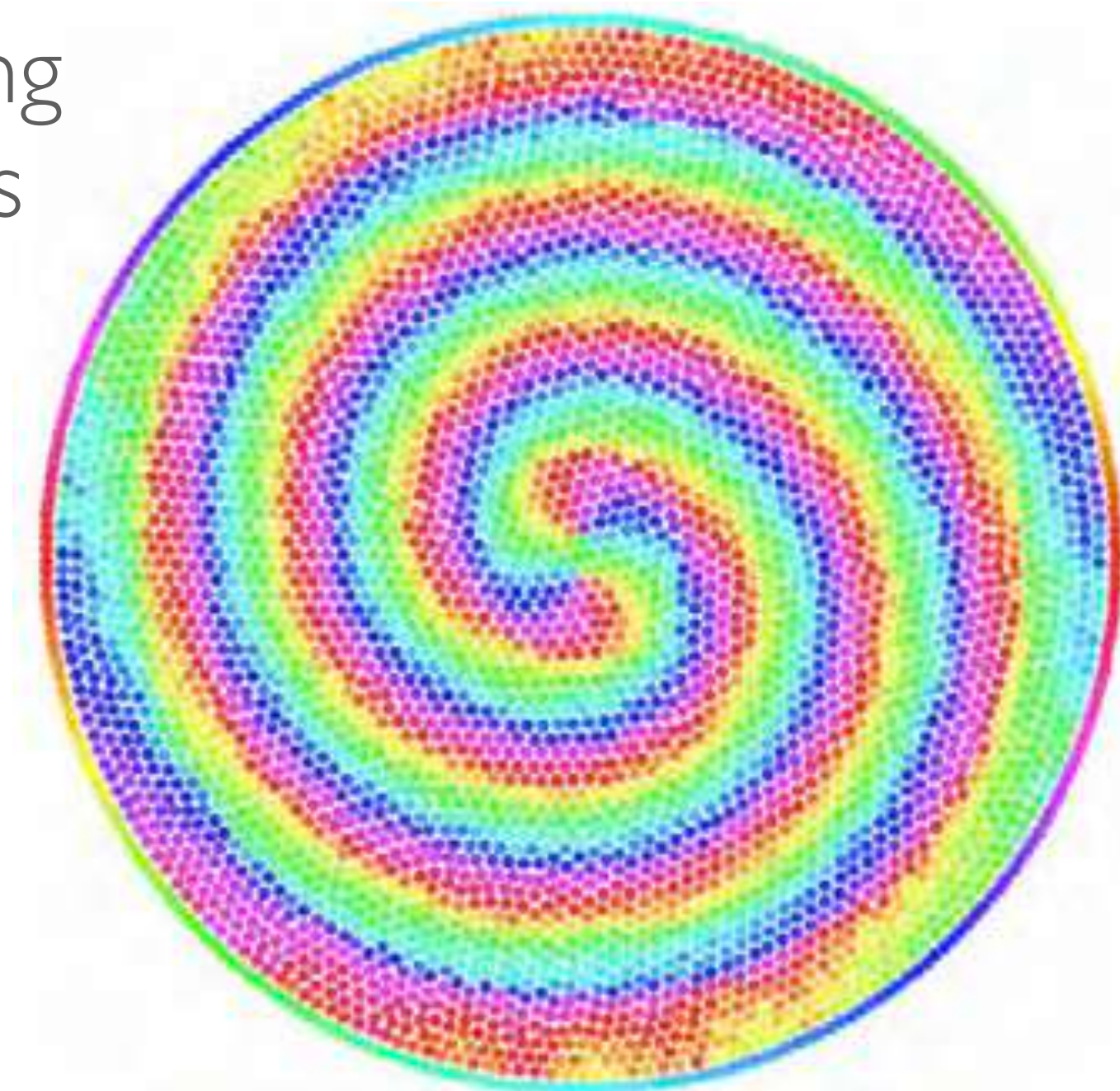
$g < 0$: anti-alignment



$k = 3$

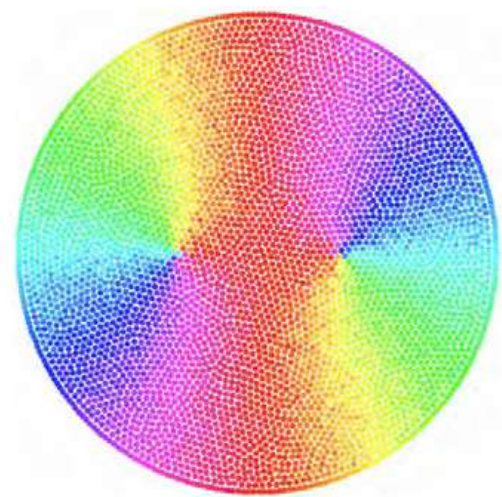


Boundary anchoring
enforces 2 defects

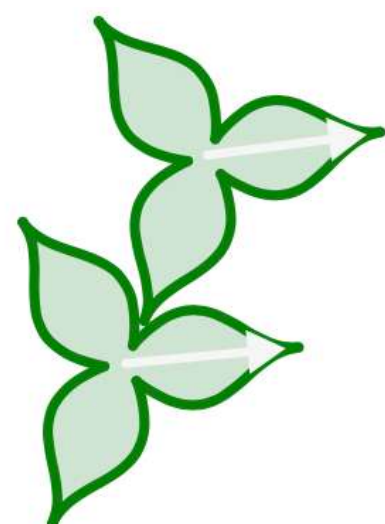


$k = 3$

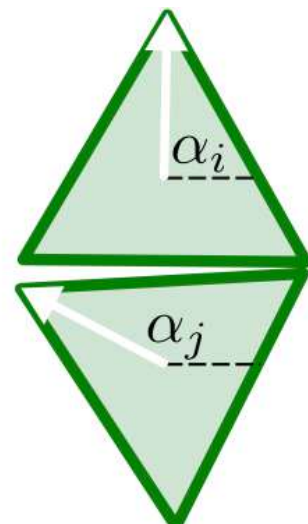
Minimal microscopic random-lattice model



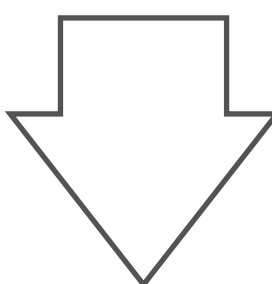
$g > 0$: alignment



$g < 0$: anti-alignment



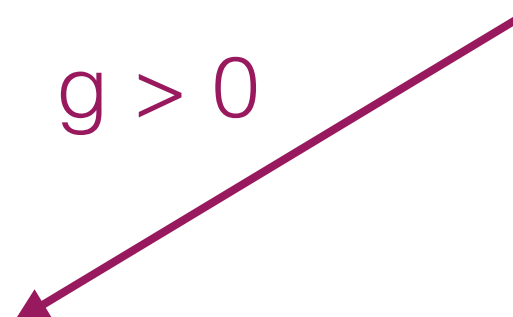
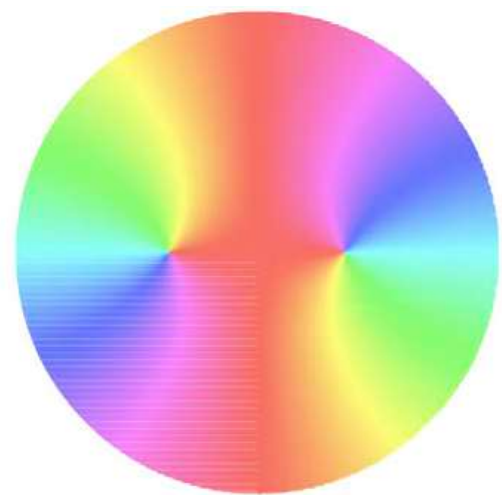
$$\frac{d\alpha_i}{dt} = \frac{g}{\pi R_\alpha^2} \sum_{j \in \mathcal{N}_i} \sin [k(\alpha_j - \alpha_i)] + \sqrt{2D_r} \xi_i(t)$$



$$\tau \partial_t \Psi_k = - \left(A + B |\Psi_k|^2 \right) \Psi_k + \mathcal{L} \left(\nabla^2 \right) \Psi_k = - \frac{\delta \mathcal{E}_k}{\delta \Psi^*}$$

Ginzburg-Landau

$$\mathcal{E}_k = \int dA \left(A |\Psi_k|^2 + B |\Psi_k|^4 + L^2 |\nabla \Psi_k|^2 \right)$$

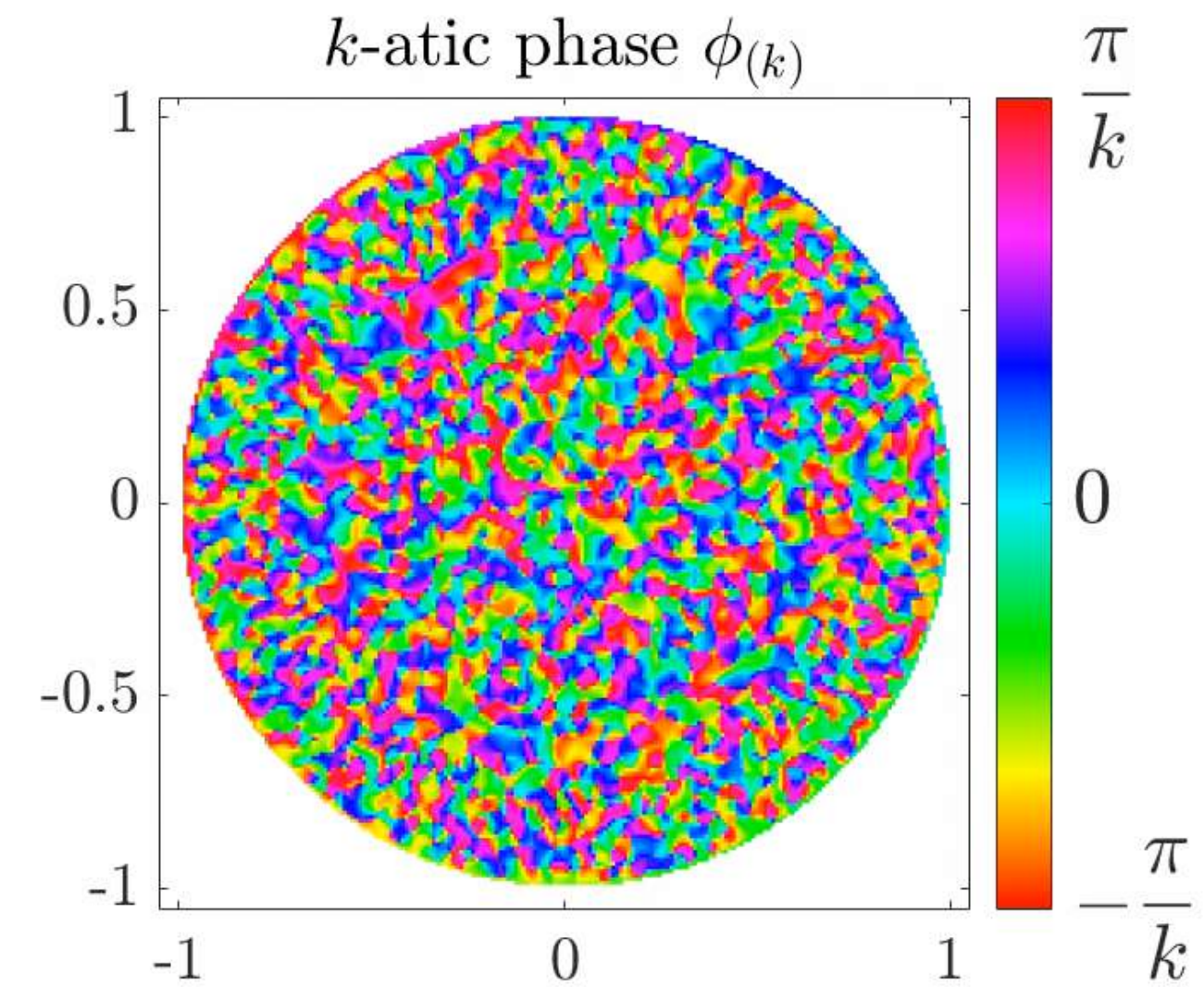
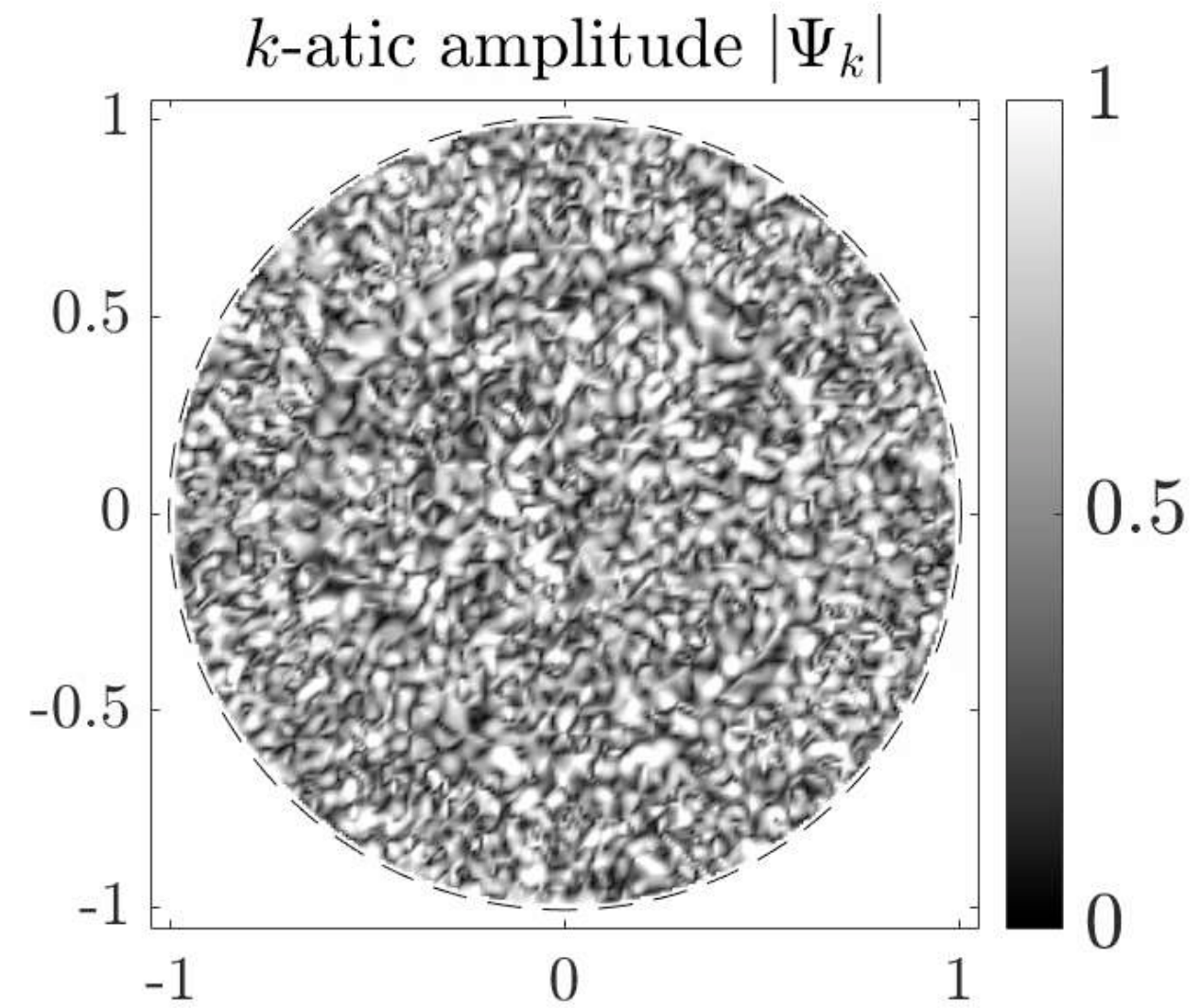
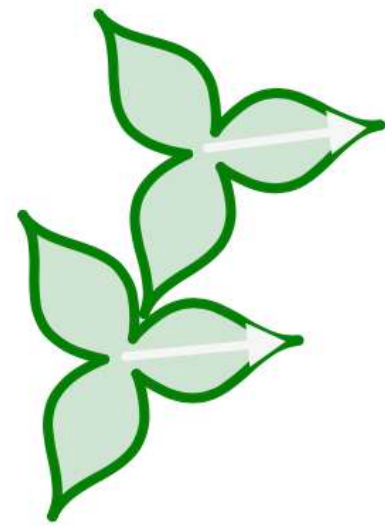


$$\mathcal{E}_k = \int dA \left(A |\Psi_k|^2 + B |\Psi_k|^4 - L_1^2 |\nabla \Psi_k|^2 + L_2^4 |\nabla^2 \Psi_k|^2 \right)$$

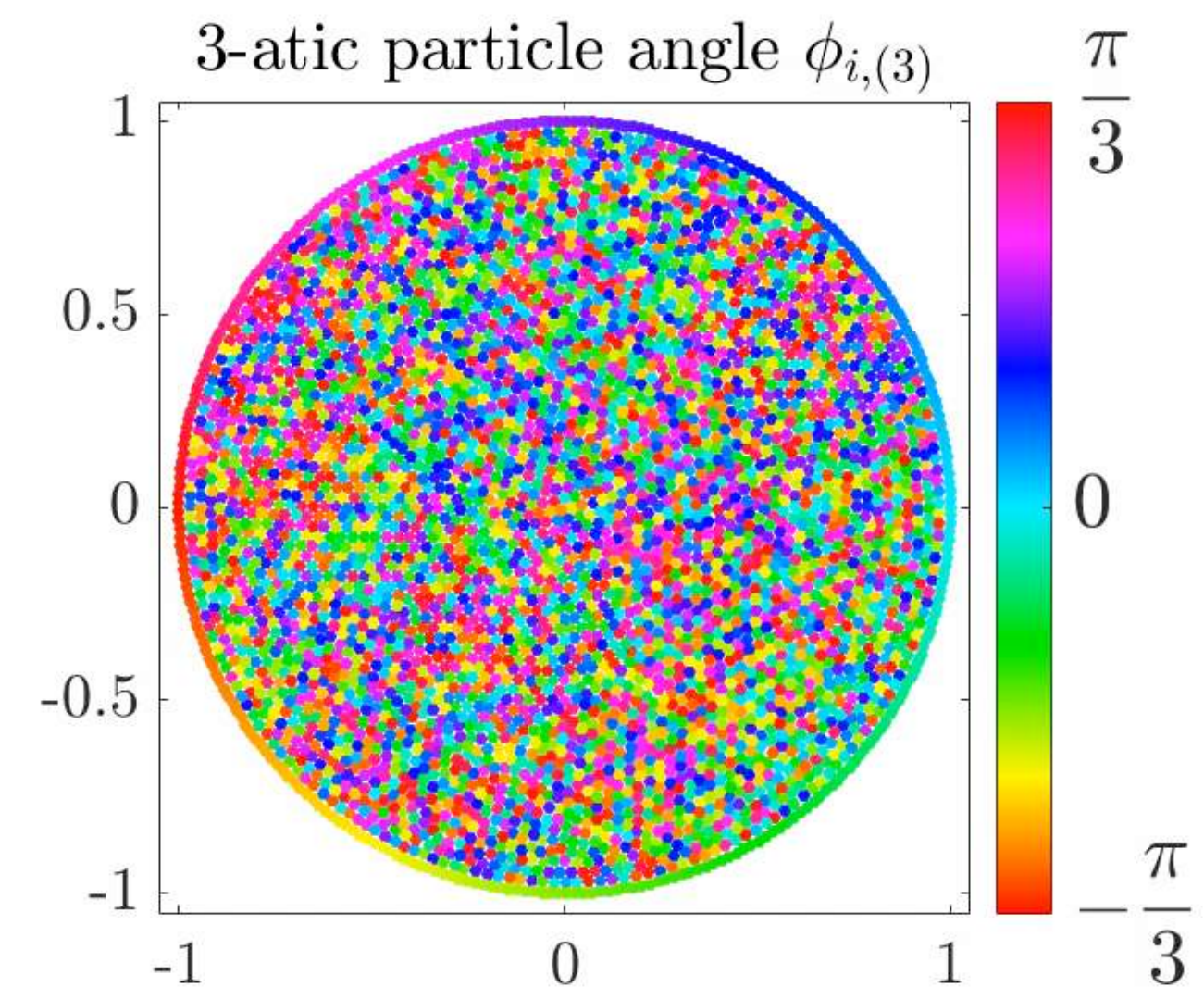
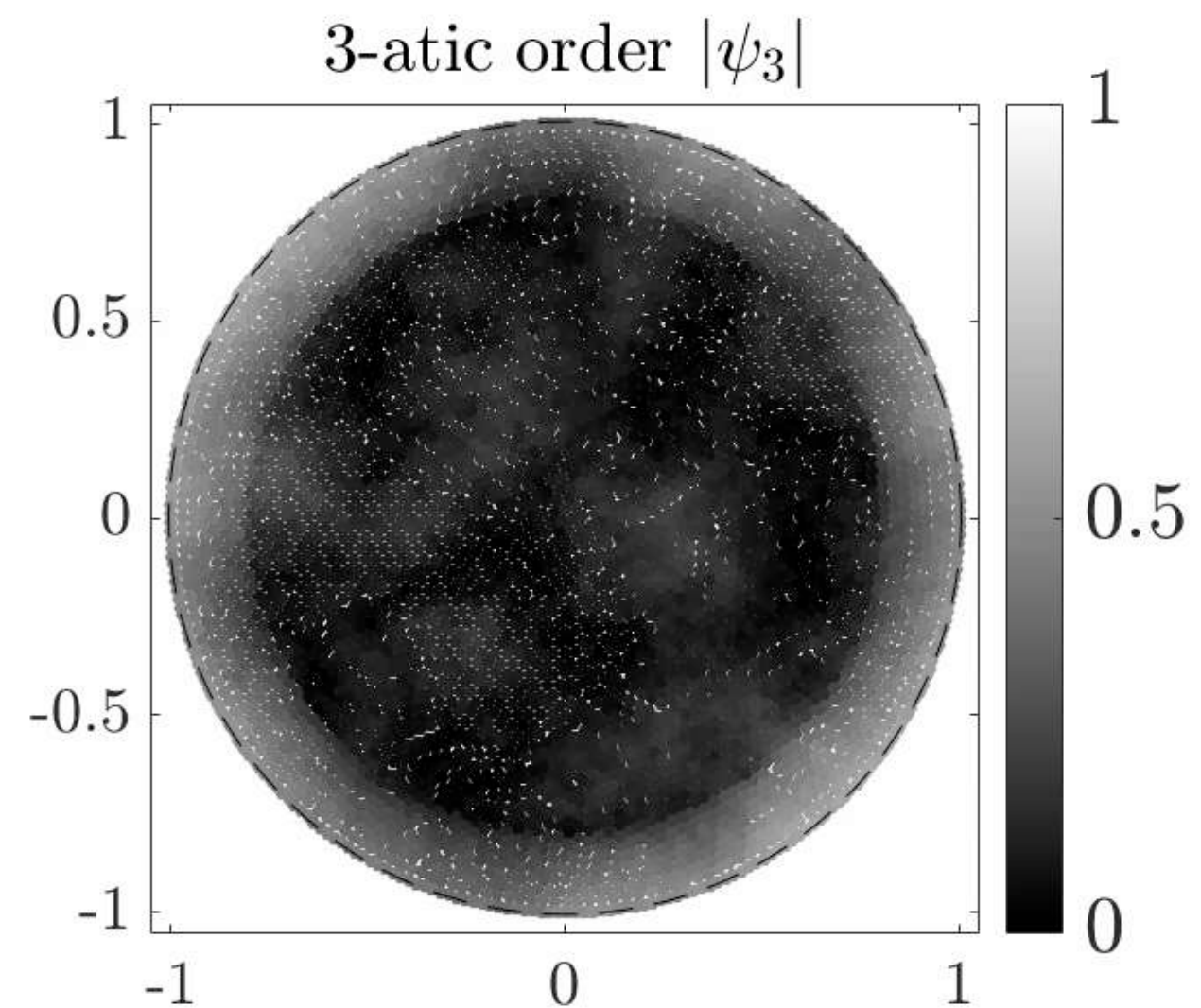
Landau-Brazovskii-Swift-Hohenberg



Relaxation dynamics

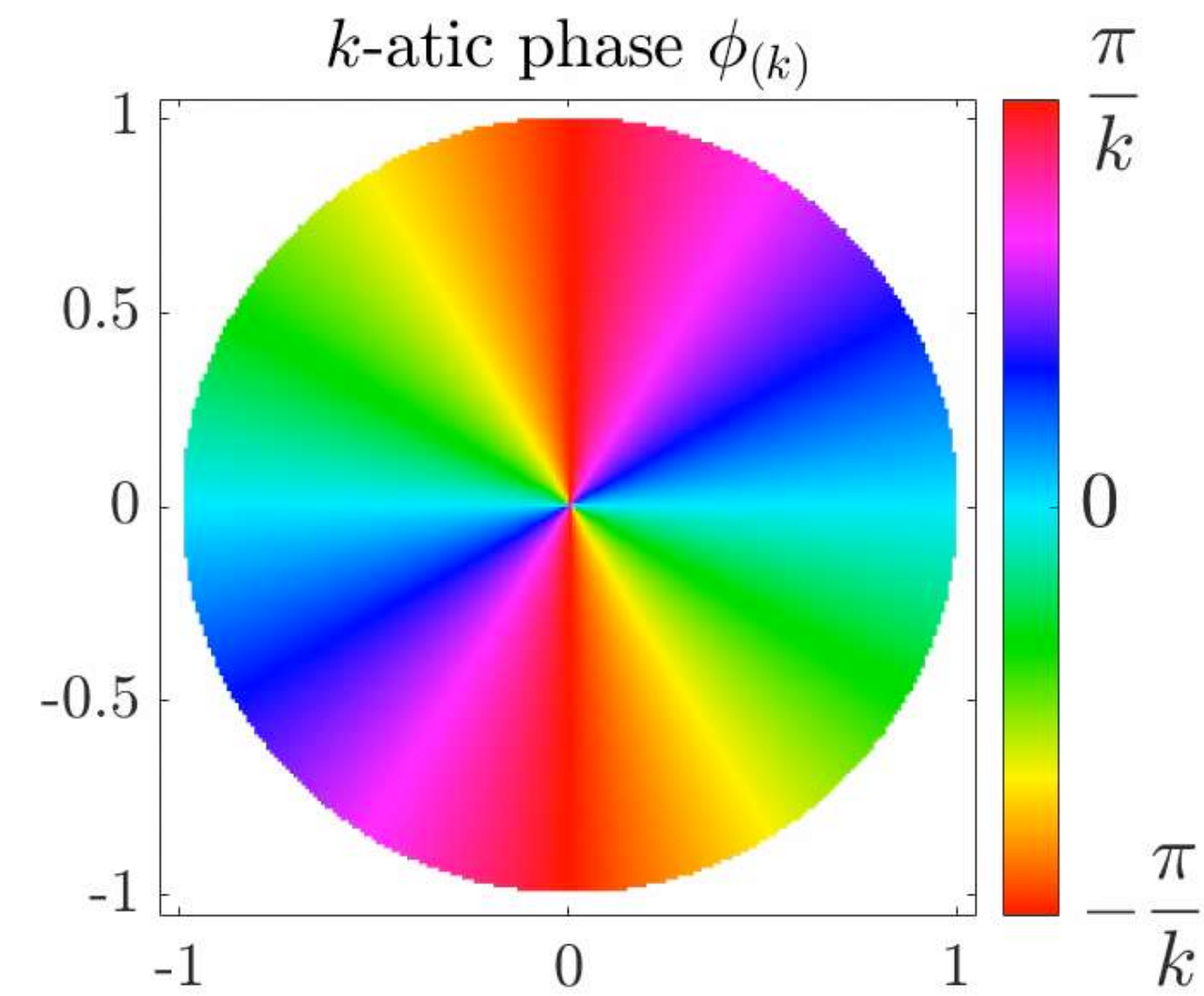
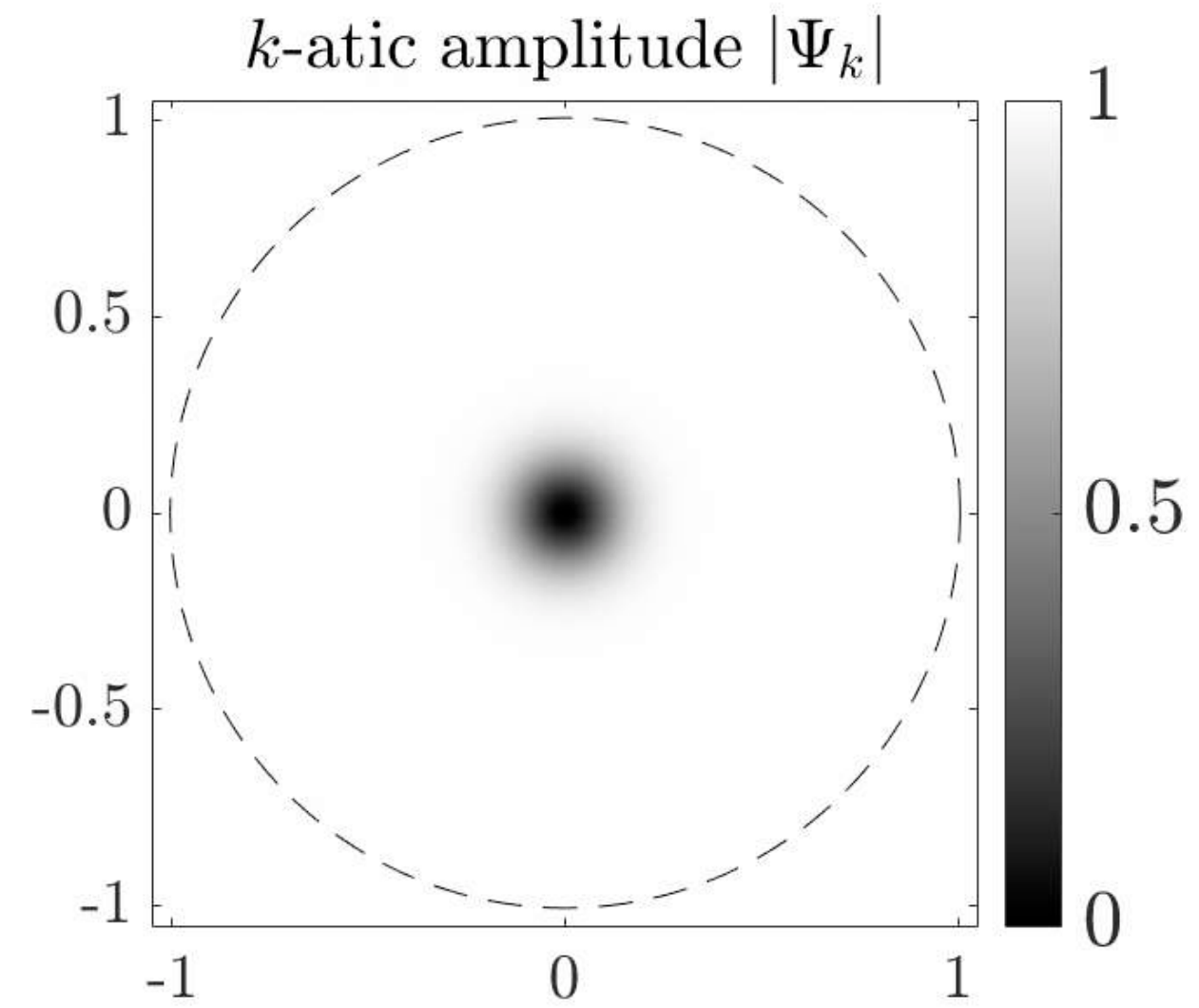
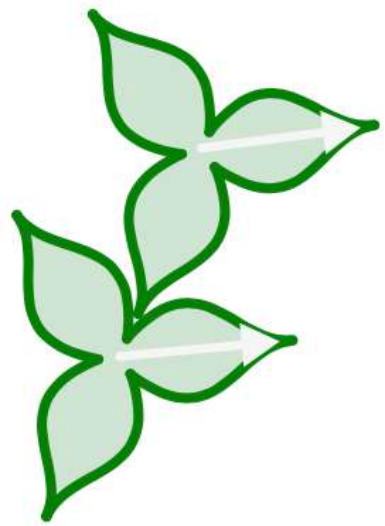


**Continuum
theory**

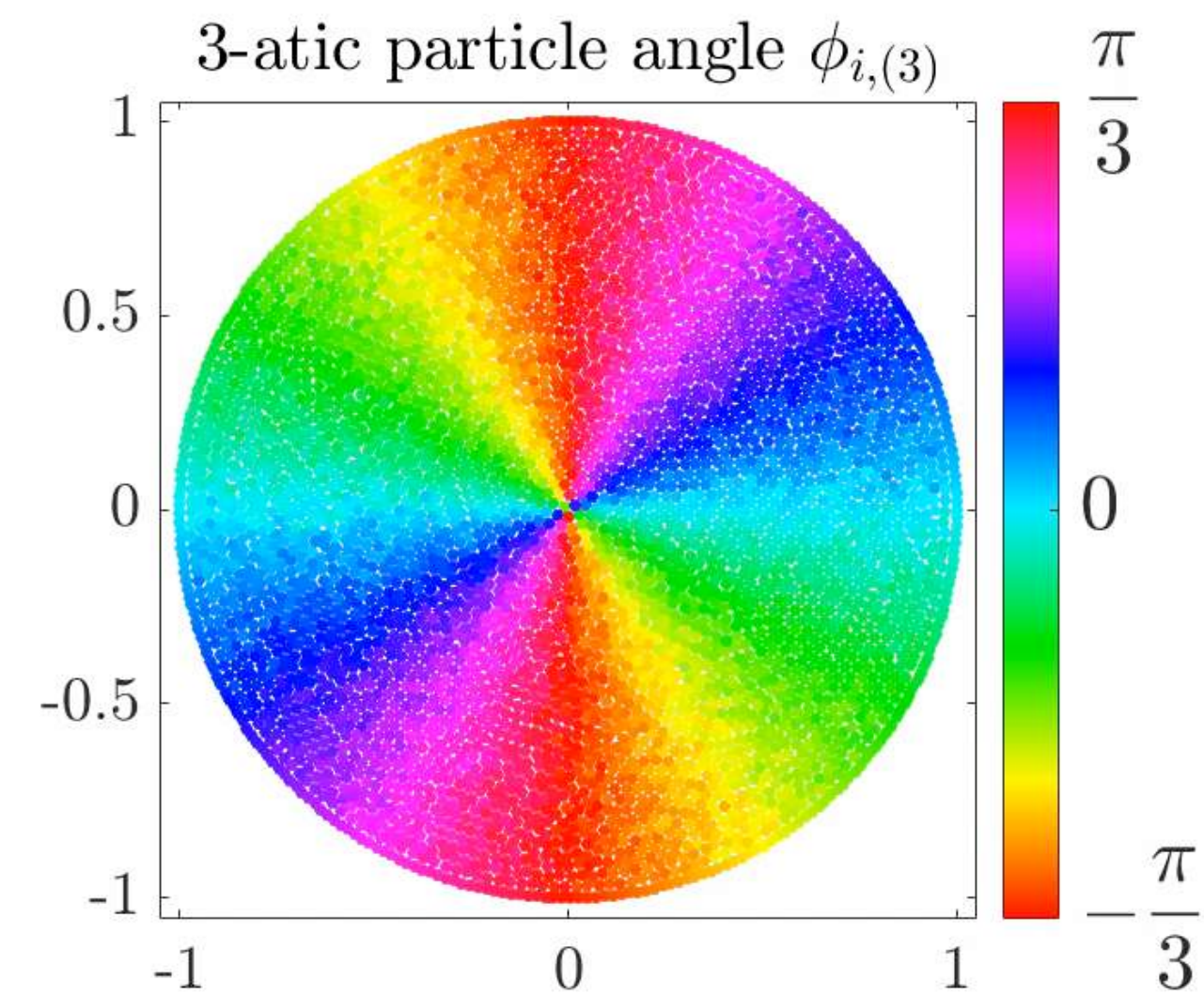
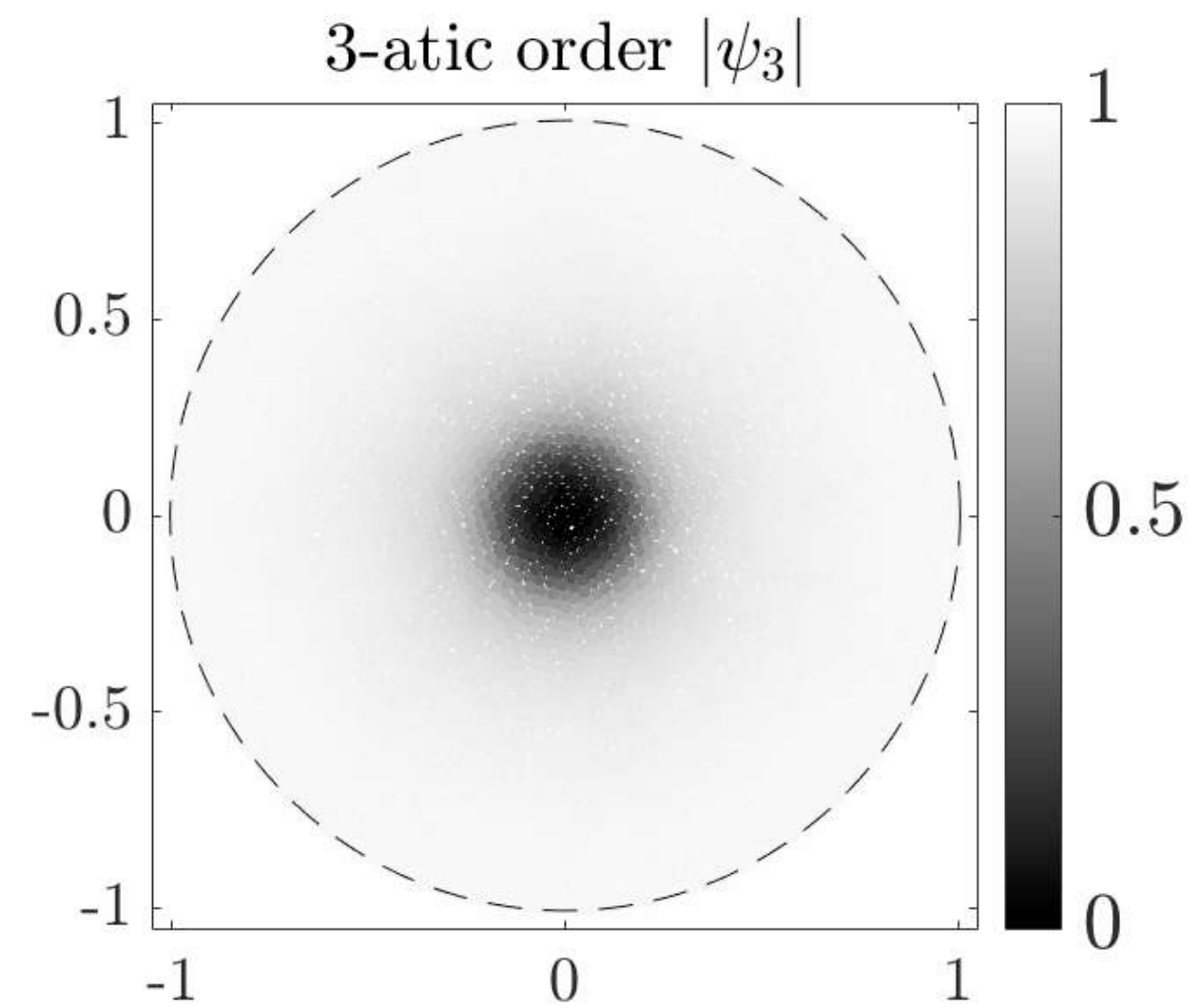


**Particle
simulation**

Defect splitting



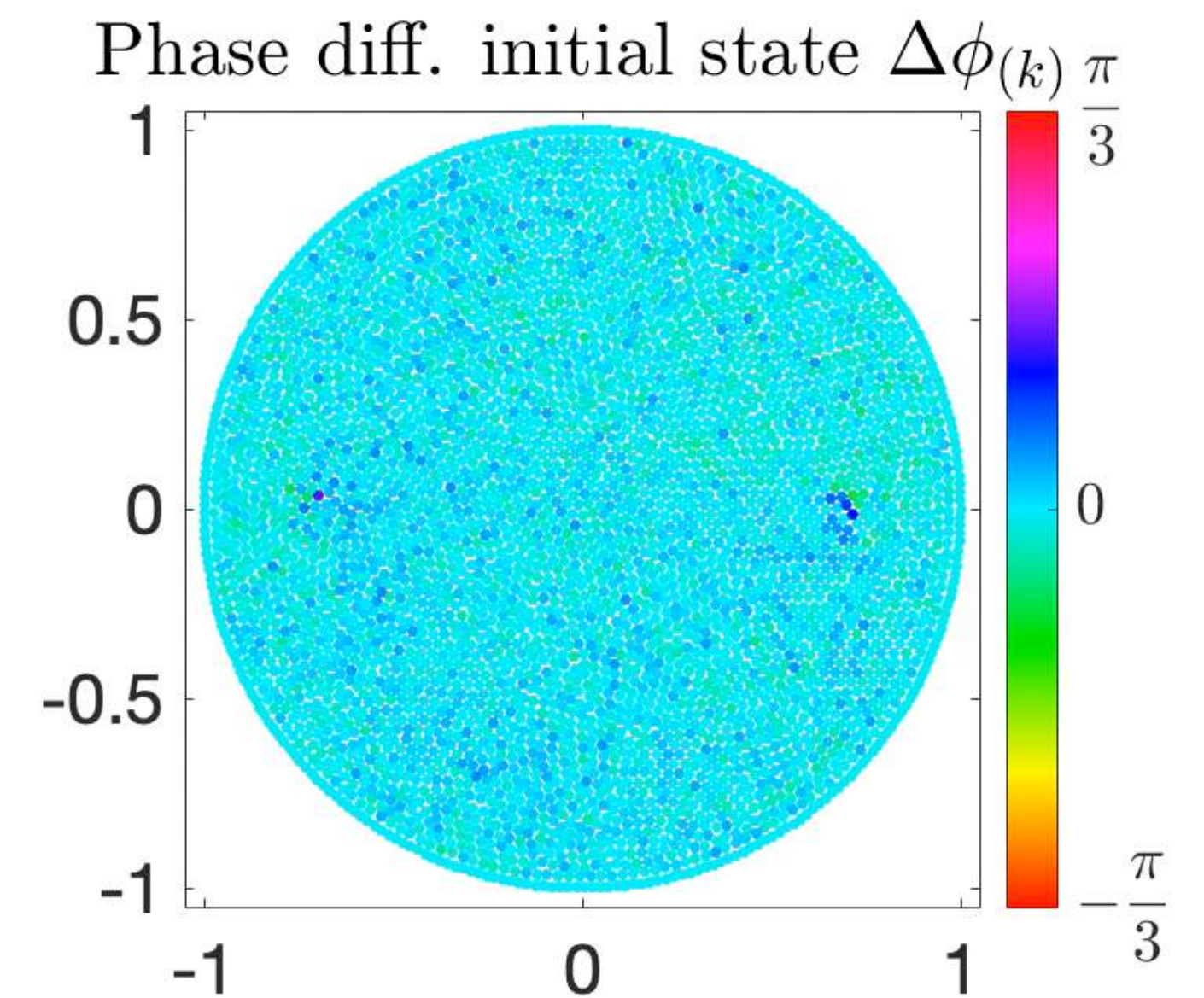
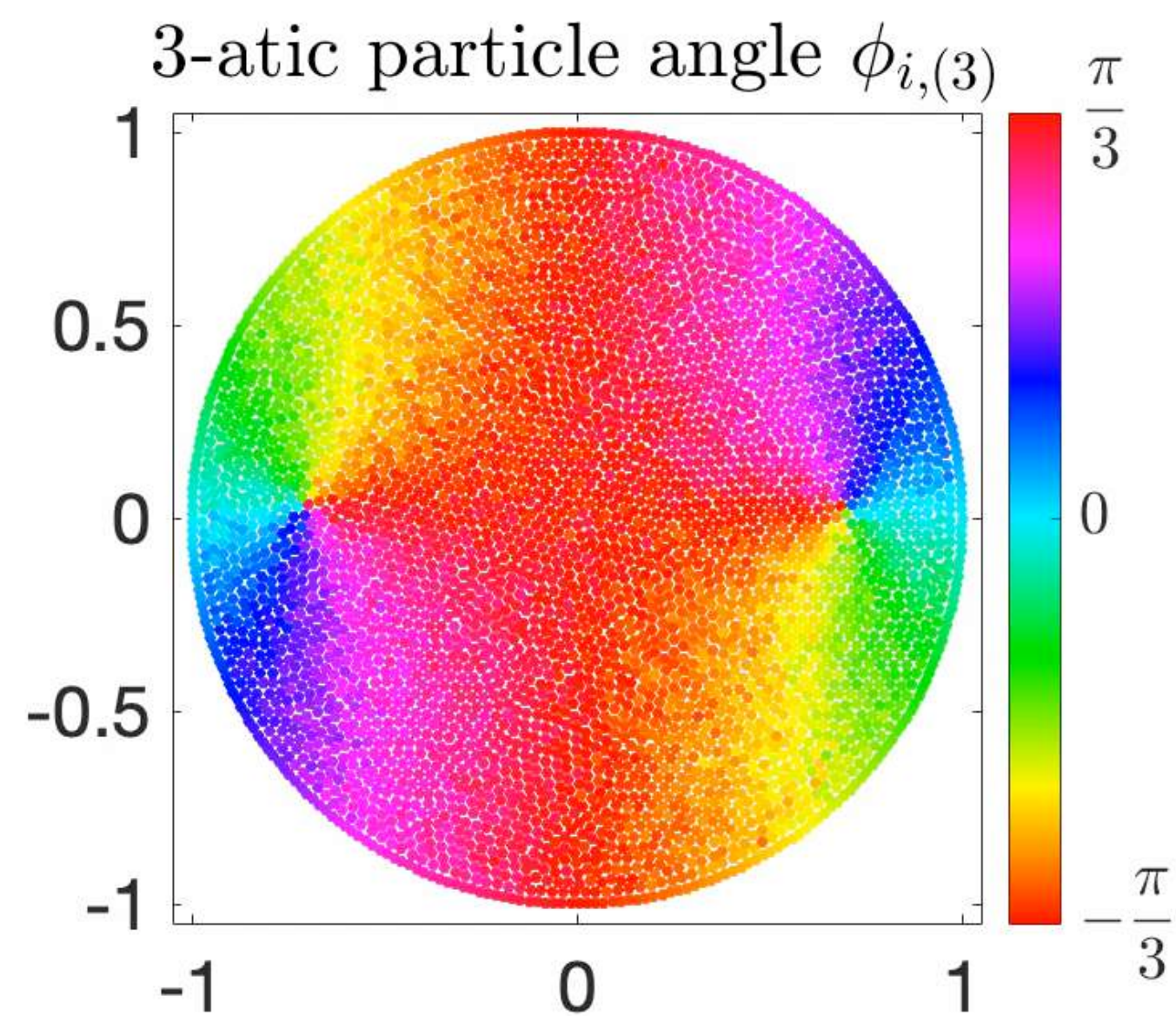
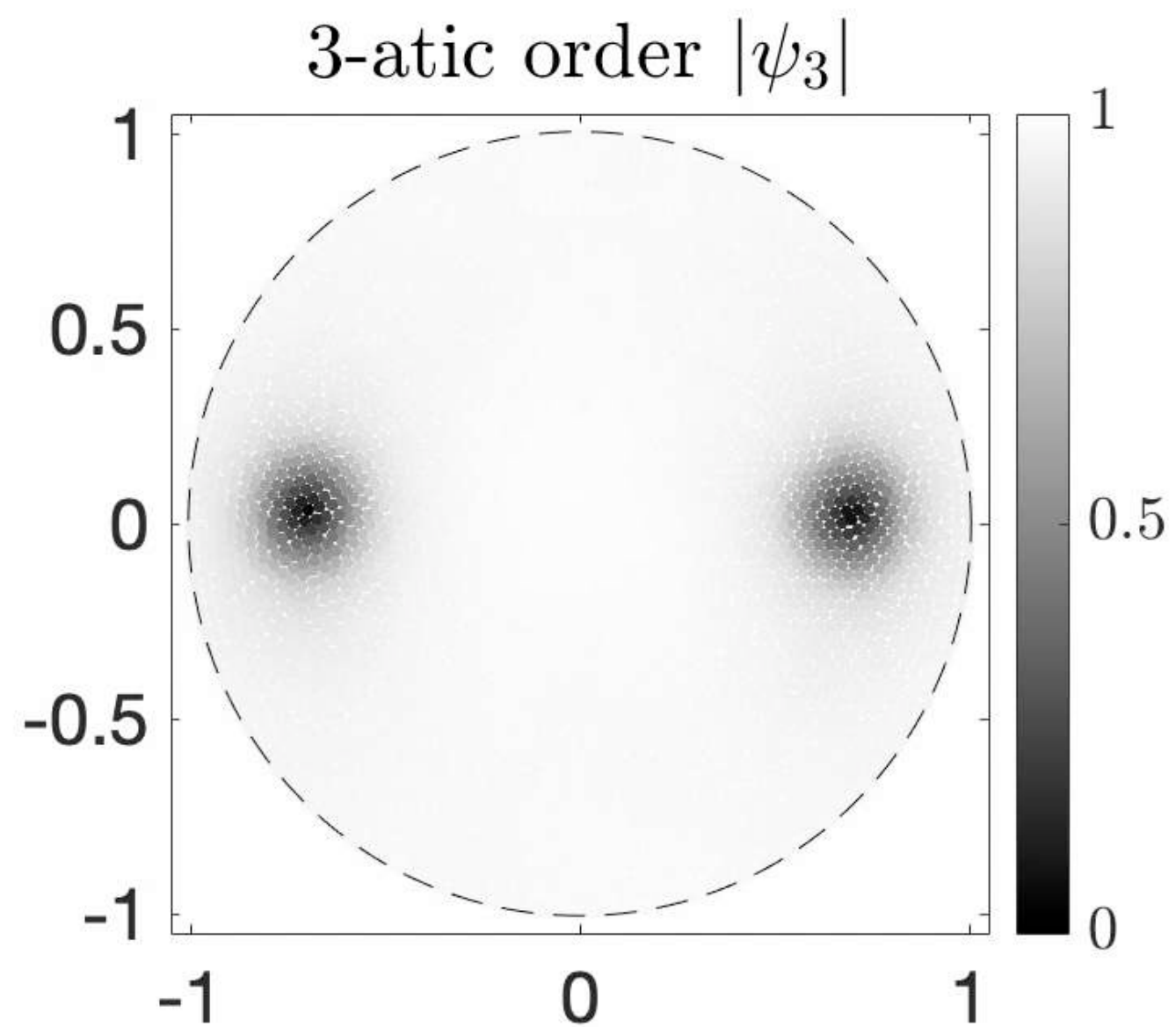
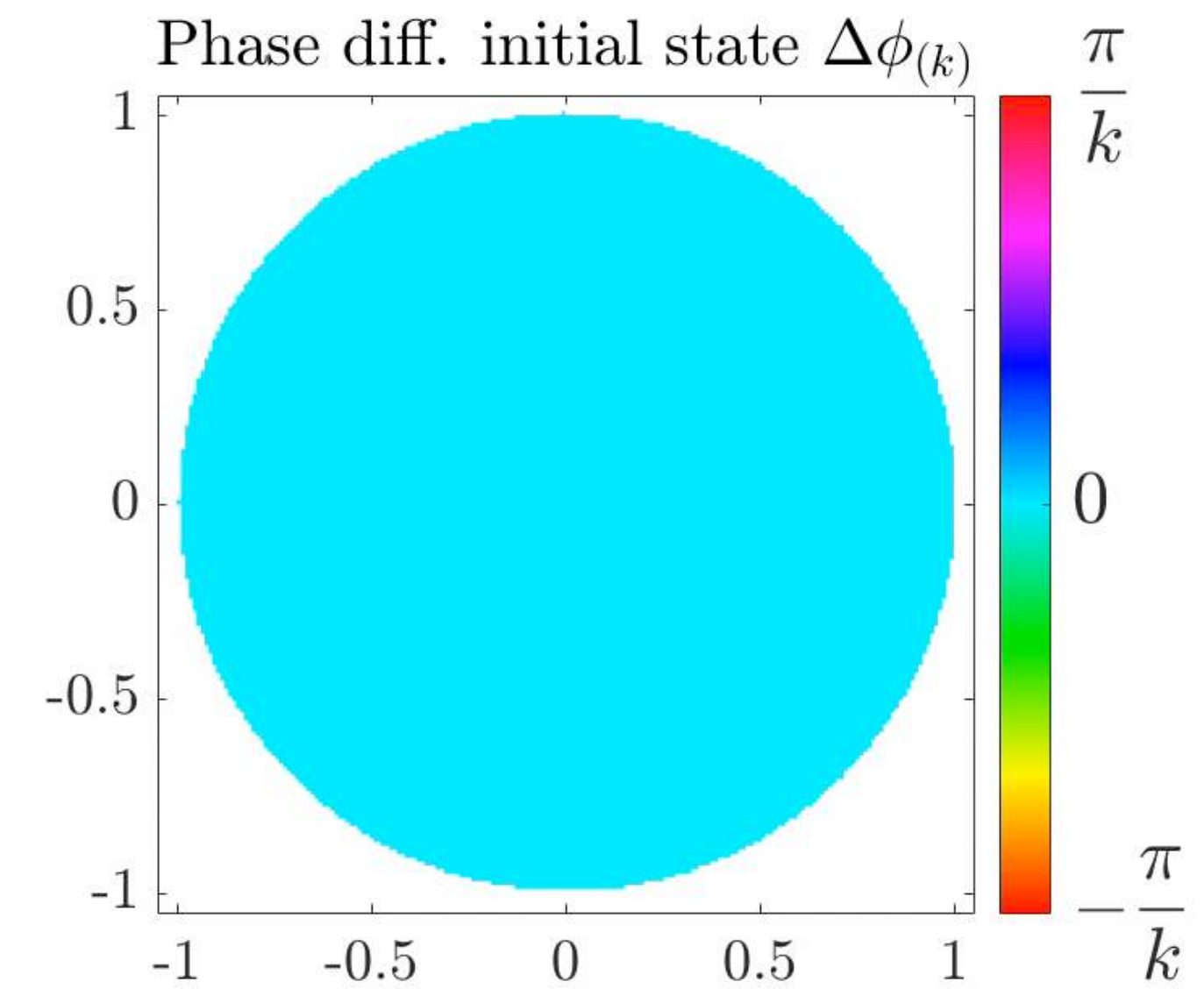
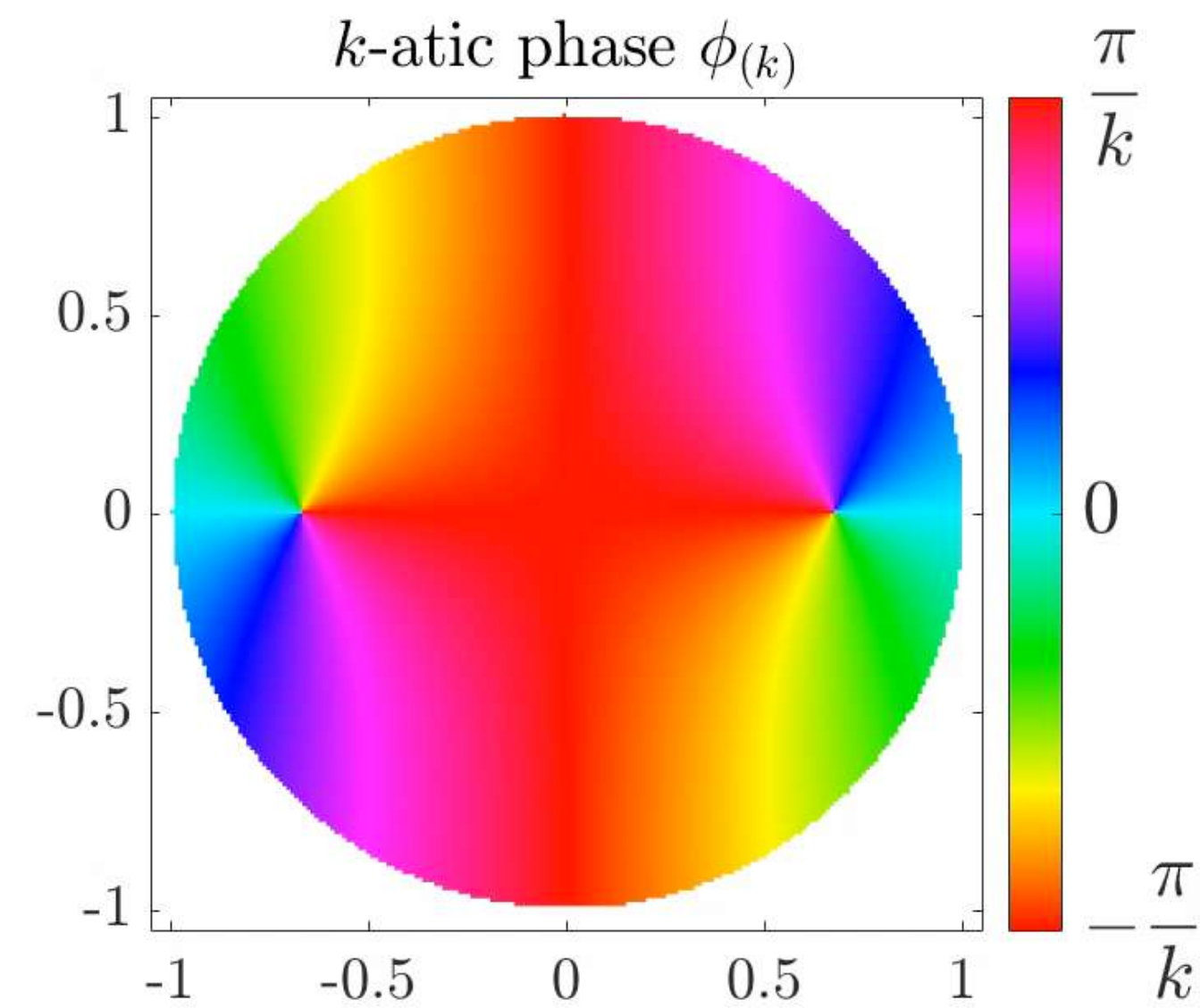
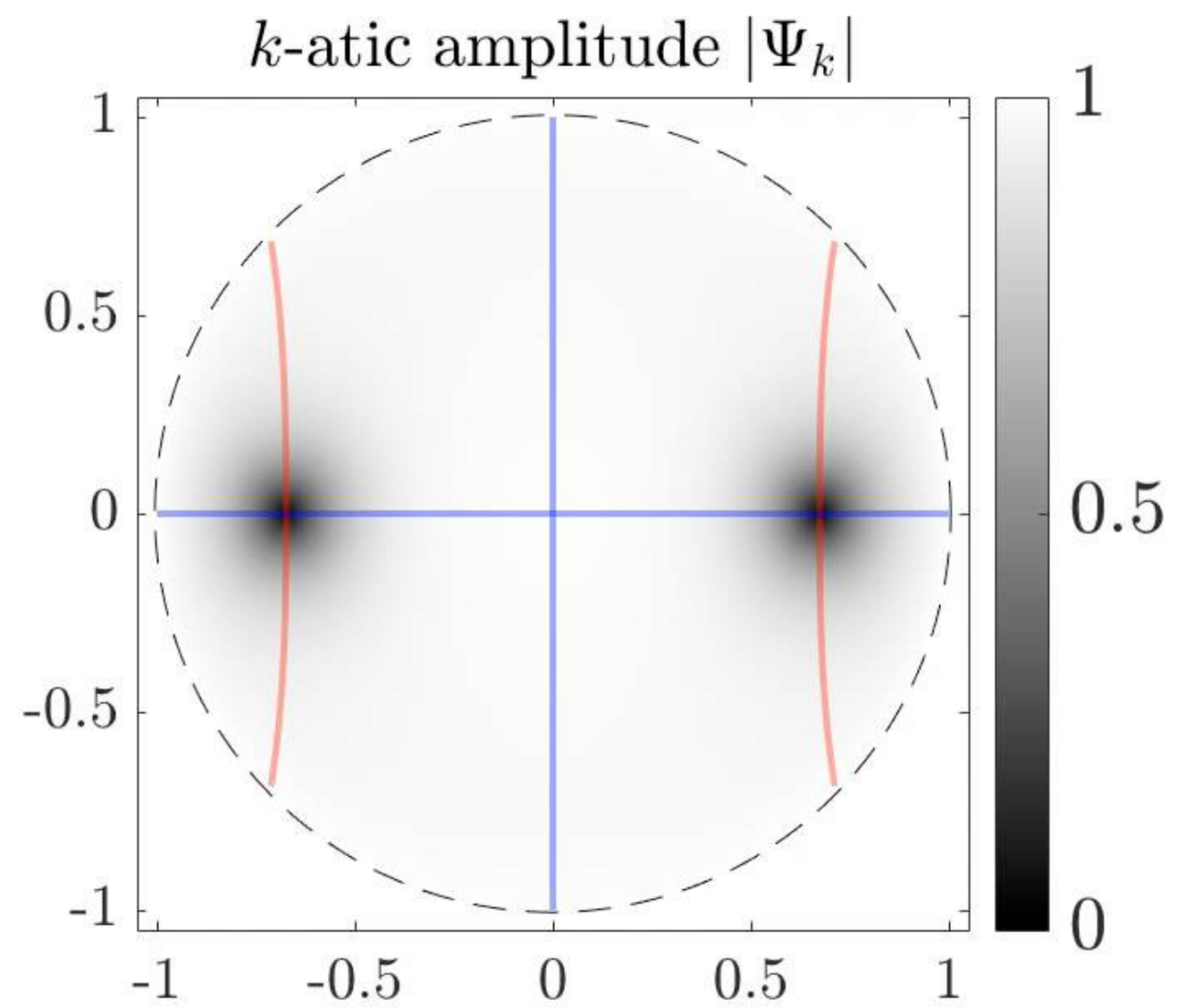
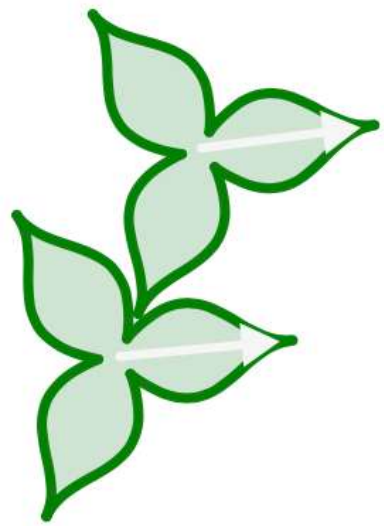
**Continuum
theory**



**Particle
simulation**

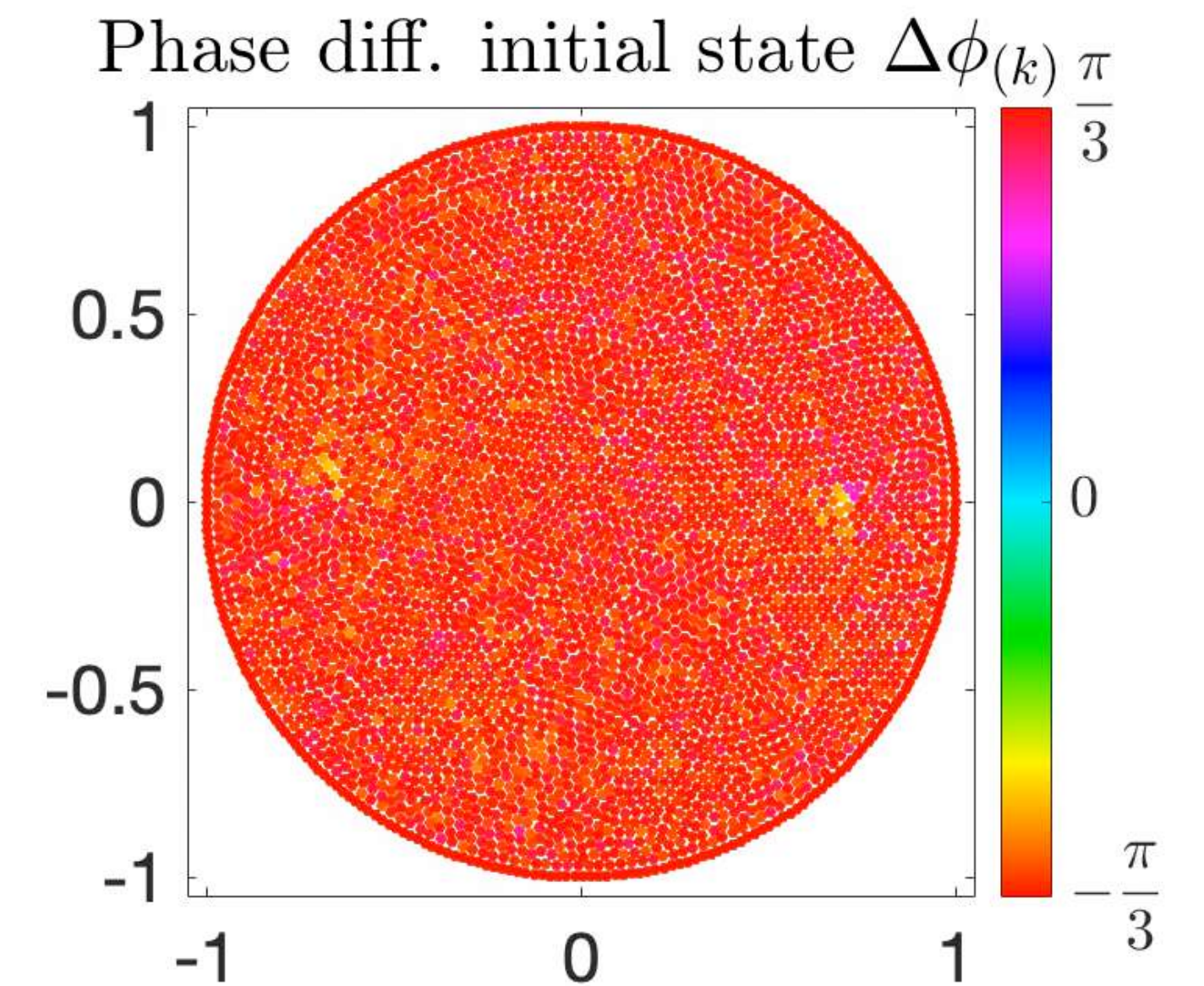
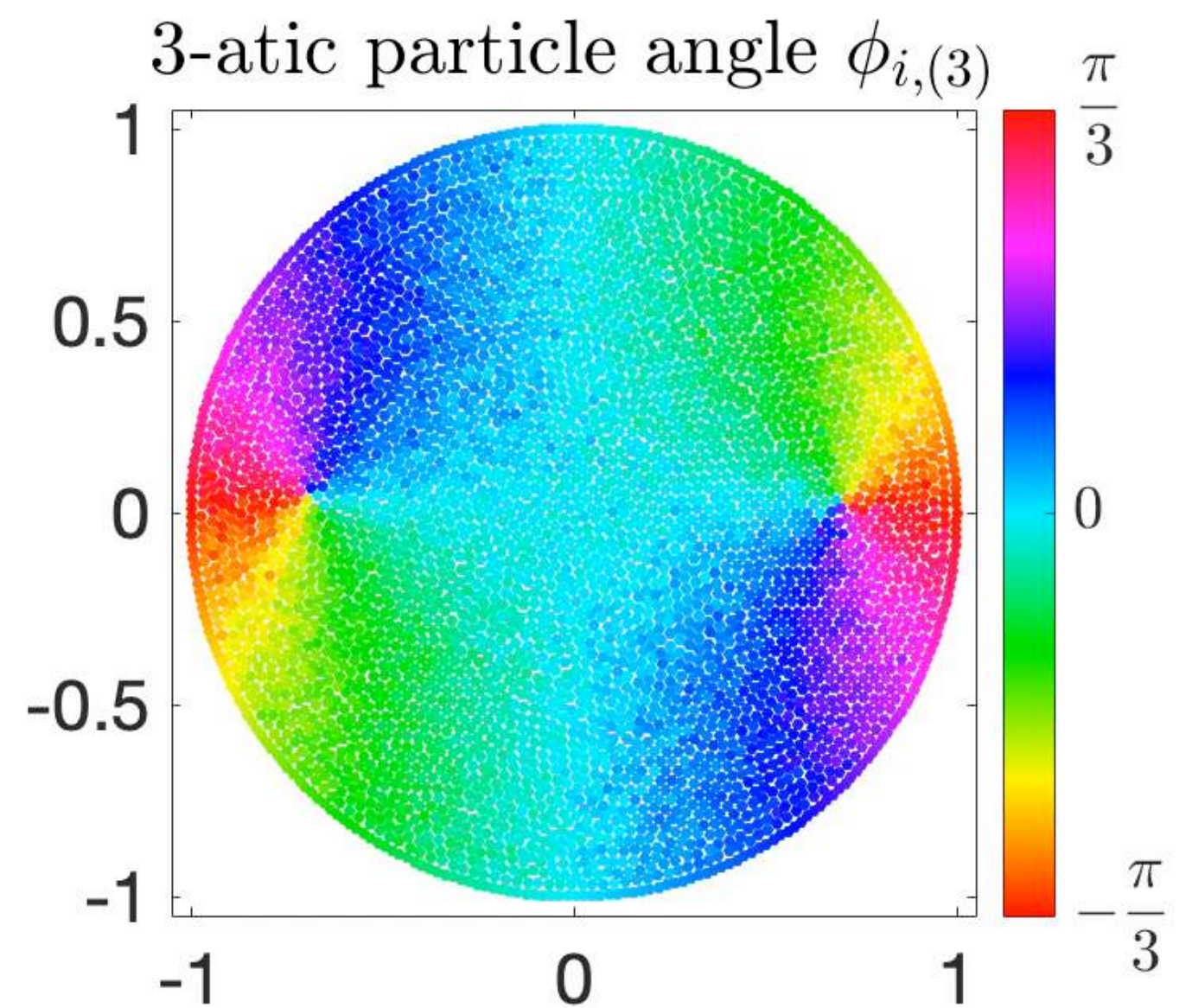
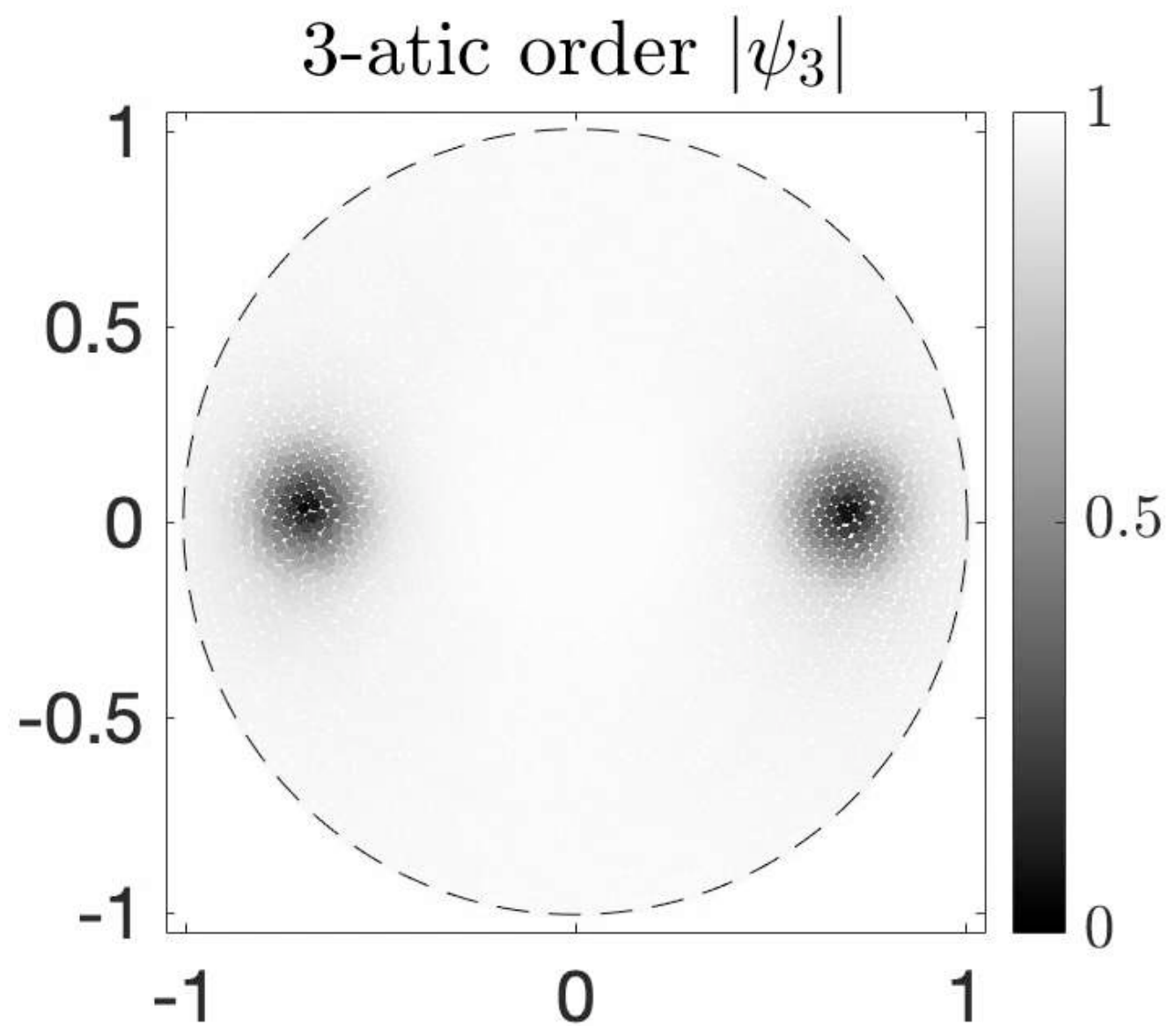
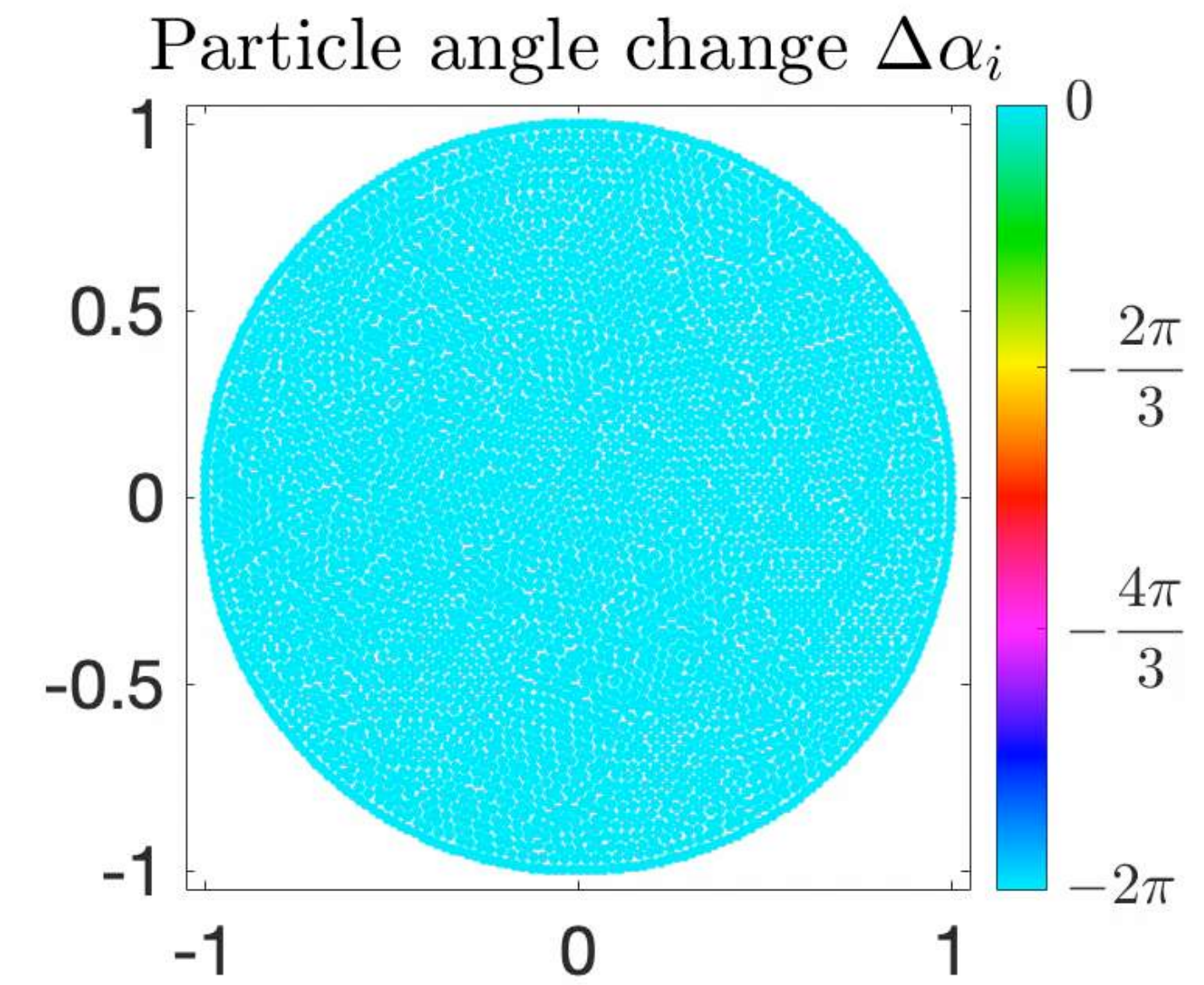
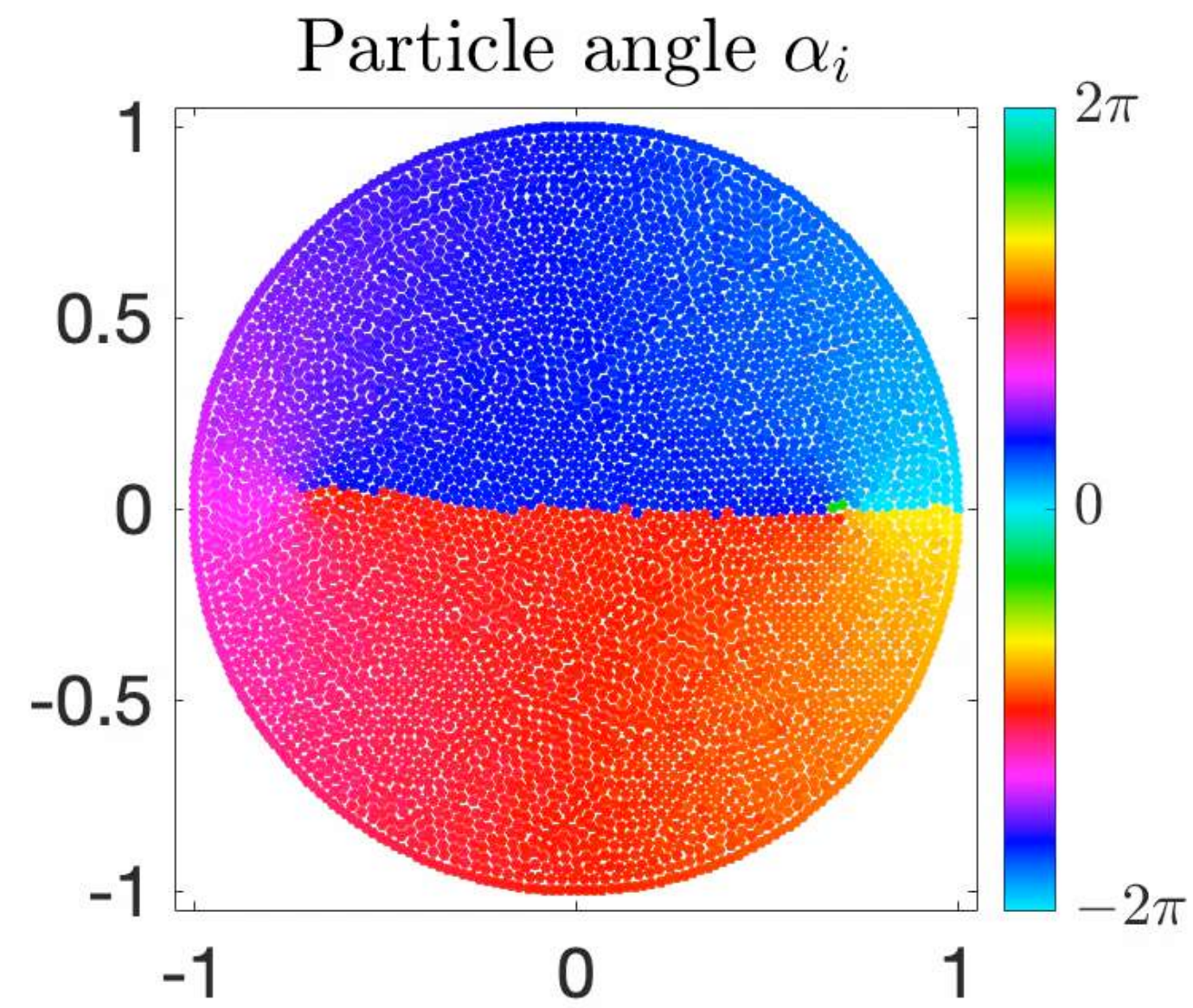
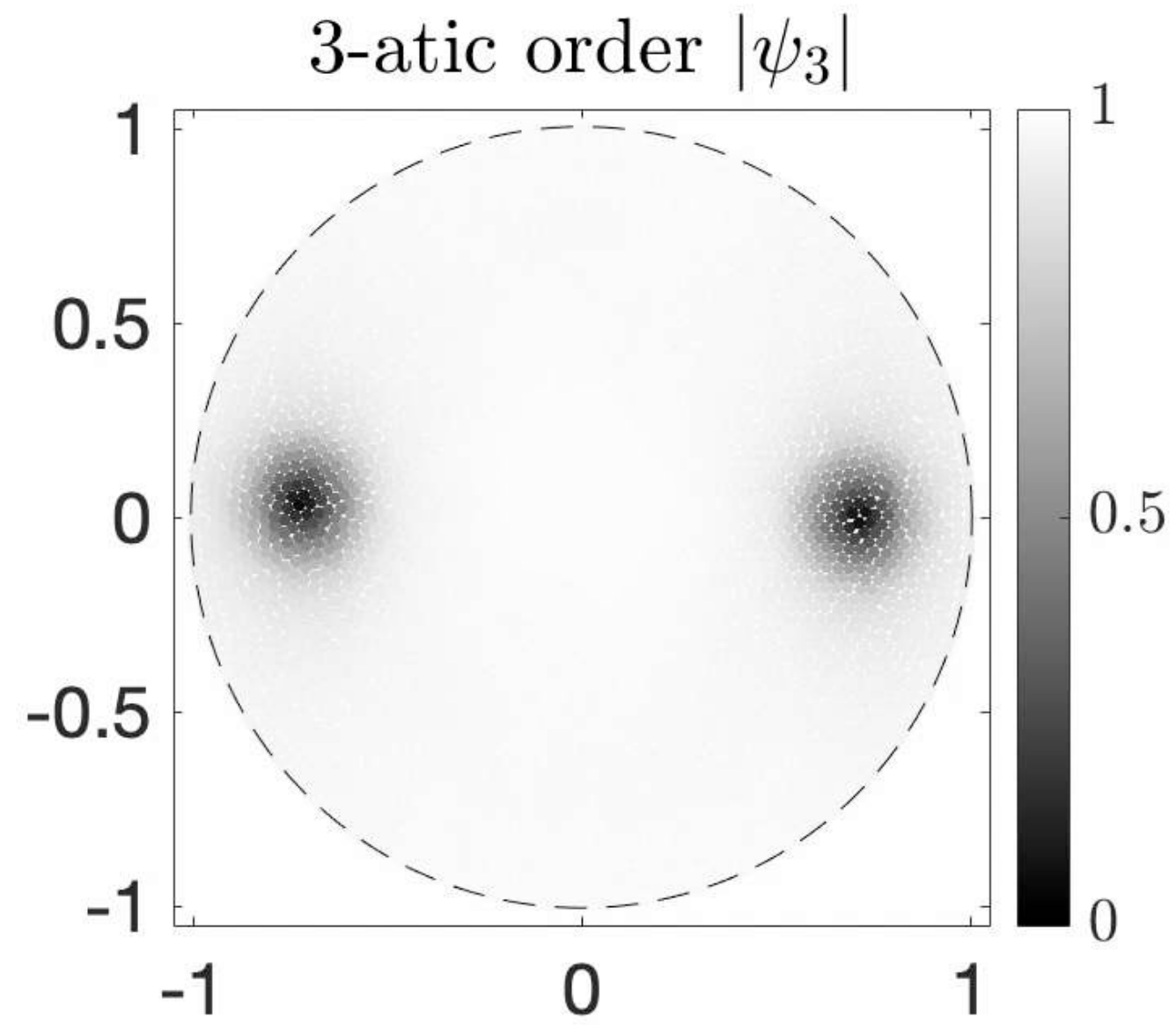
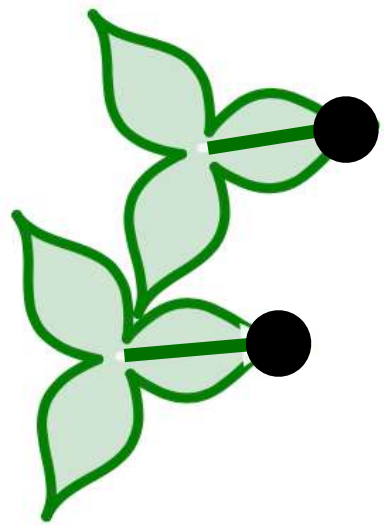
Defect braiding

Time-dependent boundary



Defect braiding

‘anyon-like’





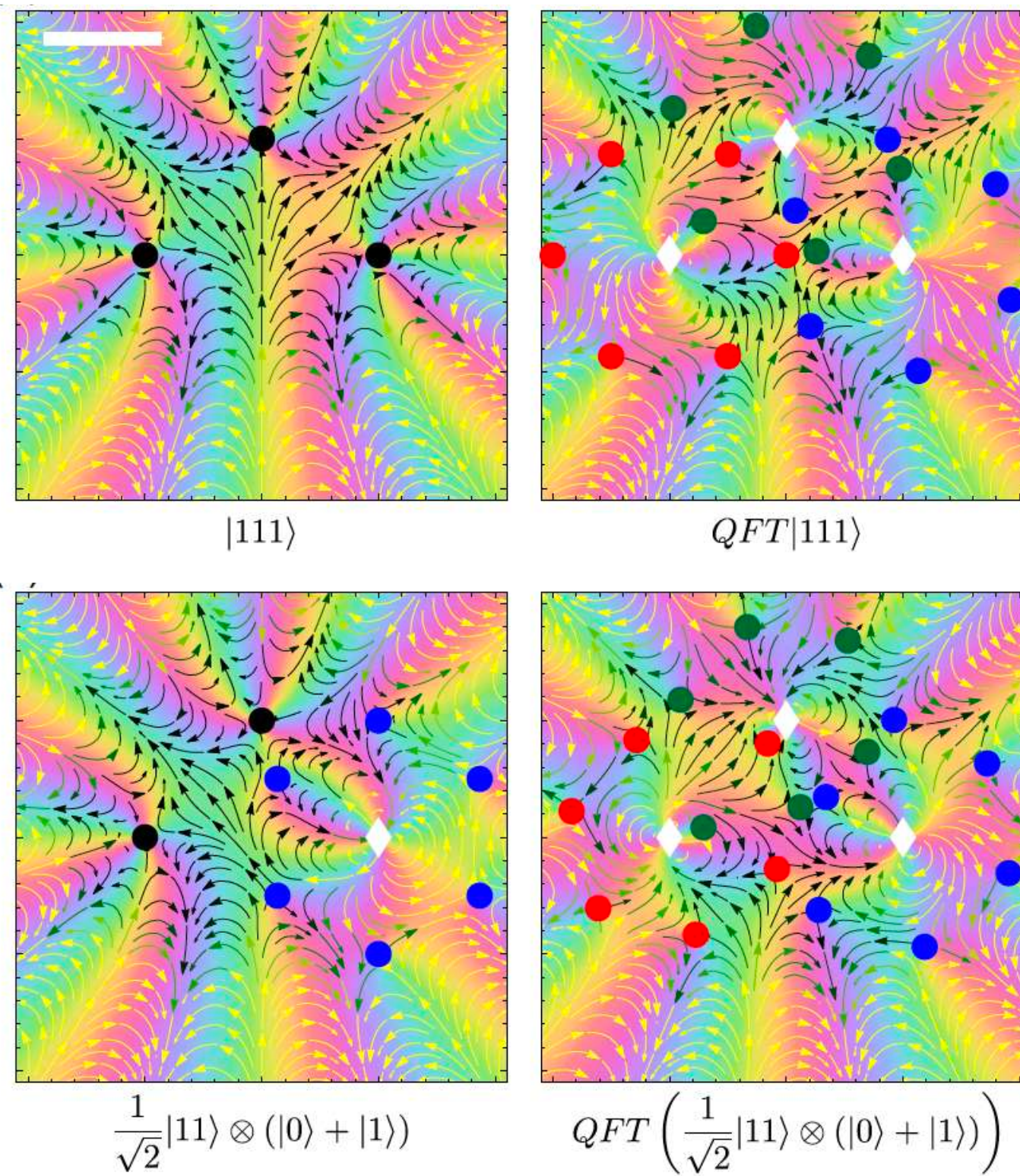
More analogies



PHYSICAL REVIEW RESEARCH **6**, 043039 (2024)

Harmonic flow-field representations of quantum bits and gates

Vishal P. Patil,^{1,2,3,*} Žiga Kos^{1,3,4,5} and Jörn Dunkel^{1,3,†}

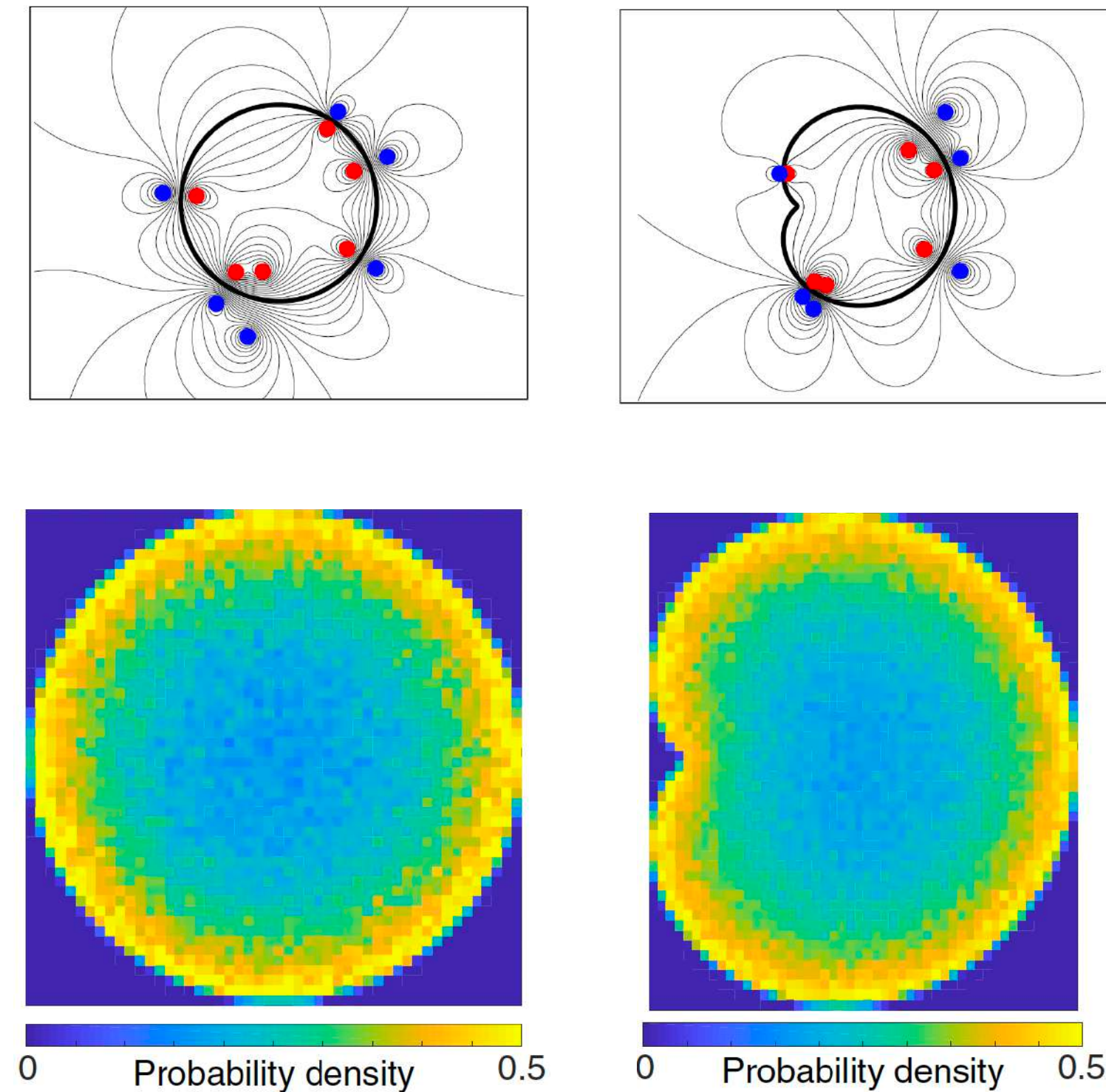


Mapping Hilbert space vectors to 2D compressible flow patterns

PHYSICAL REVIEW FLUIDS **6**, 064702 (2021)

Chiral edge modes in Helmholtz-Onsager vortex systems

Vishal P. Patil and Jörn Dunkel¹



Real space argument: edge modes from boundary conditions