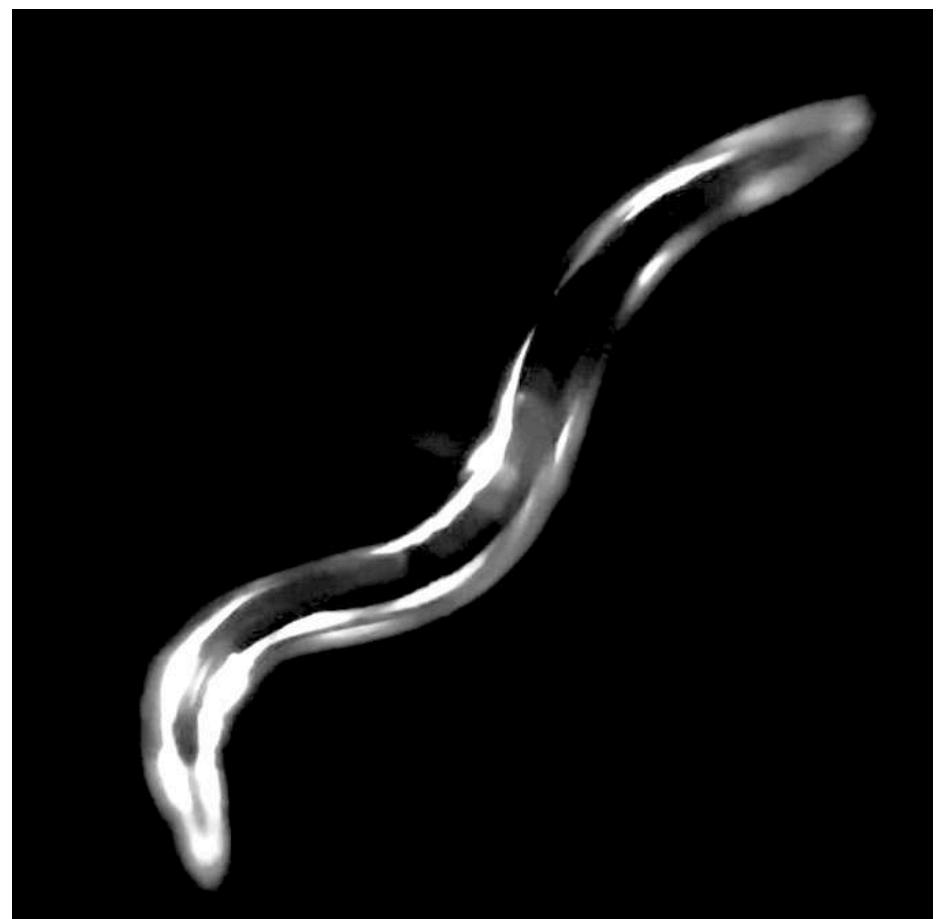
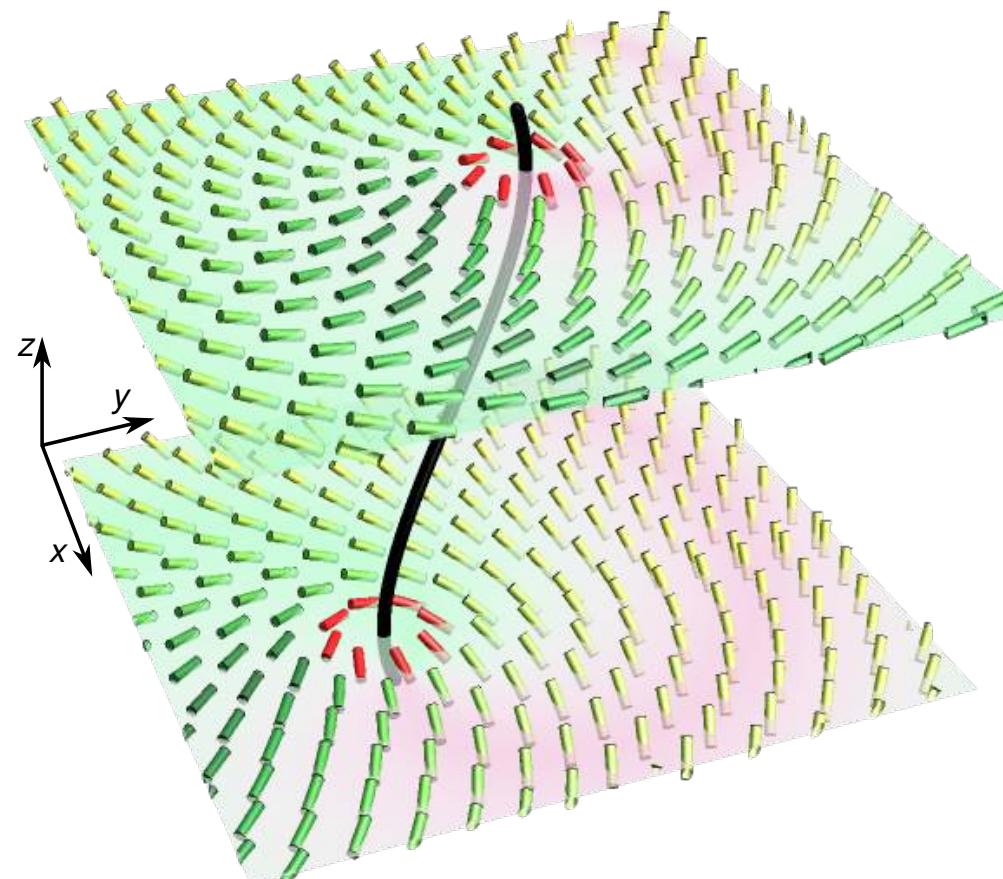


Quantum analogues in living and soft matter

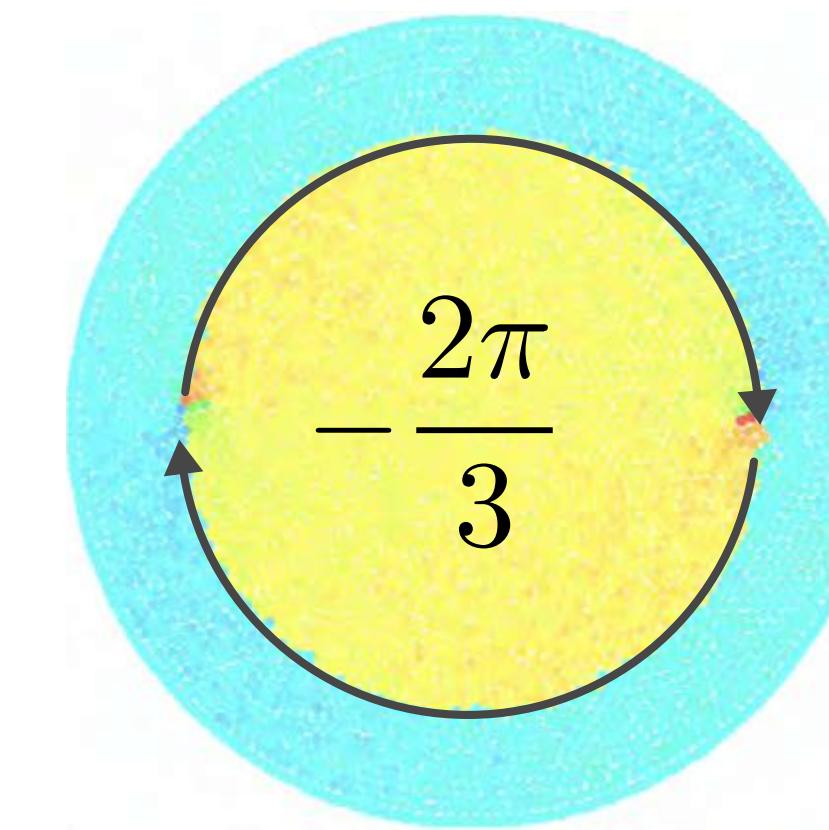


Flavell lab (MIT BCS)

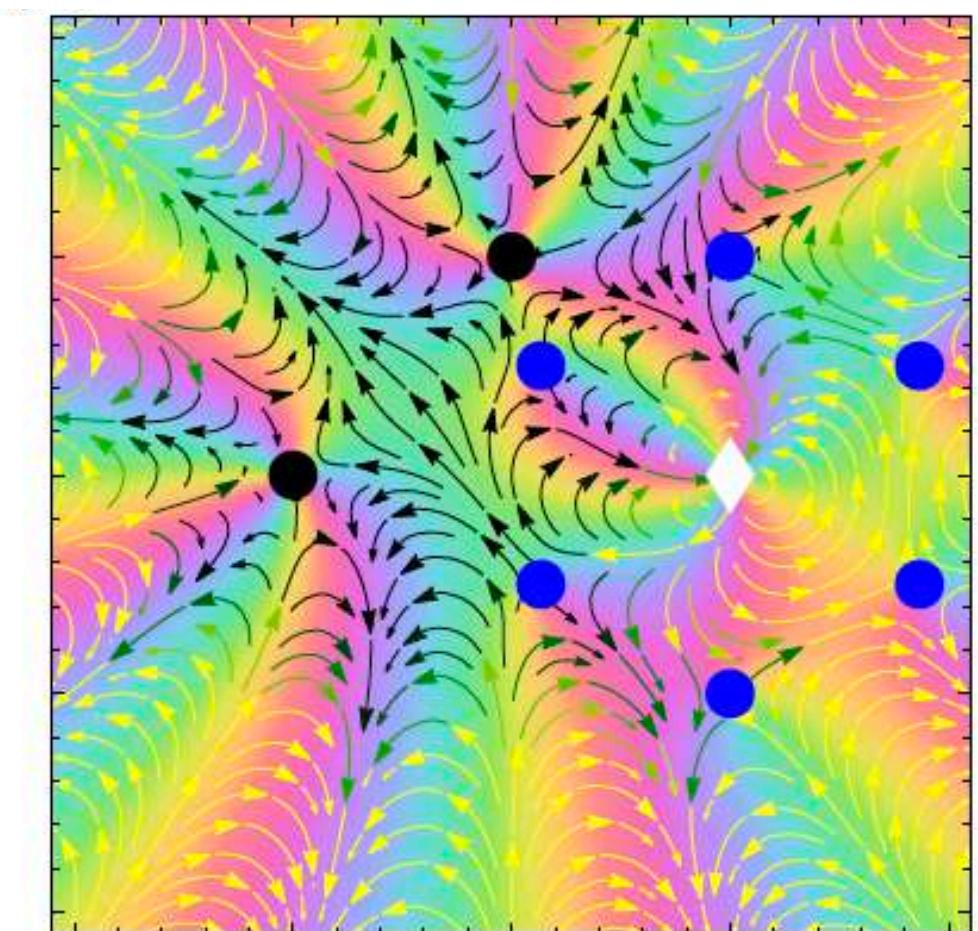


PRL 2023

Science Advances 2022



PRX 2022



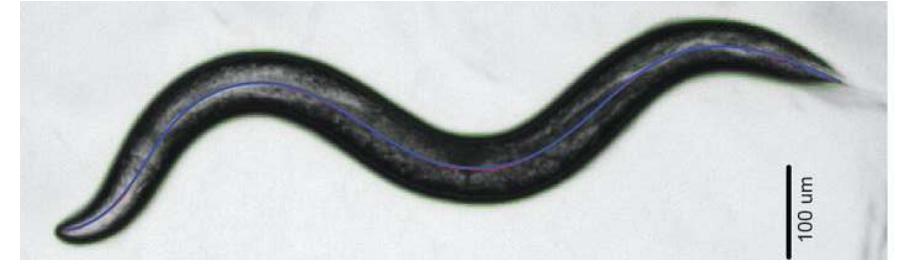
PRR 2024



Outline

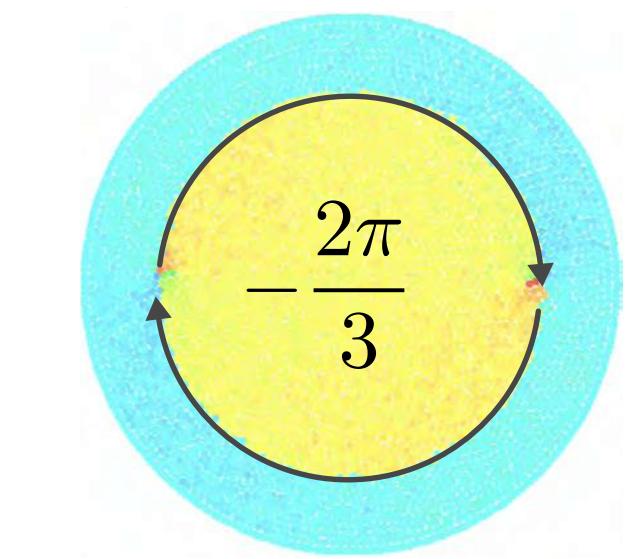
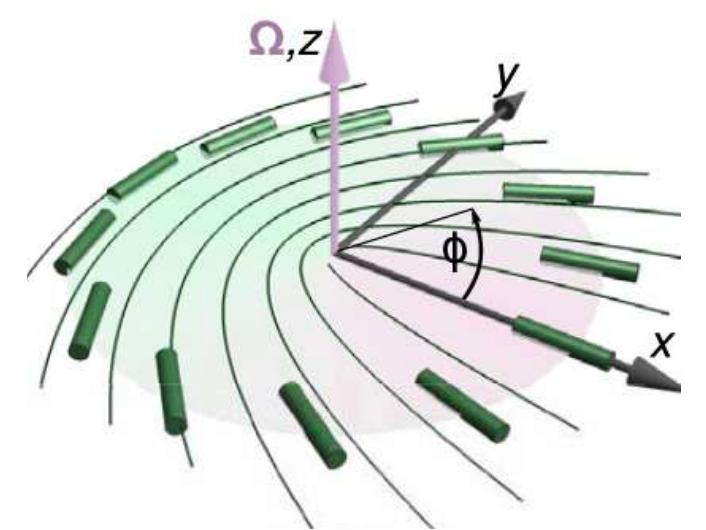
1. Dynamics of quasi-1D living systems

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



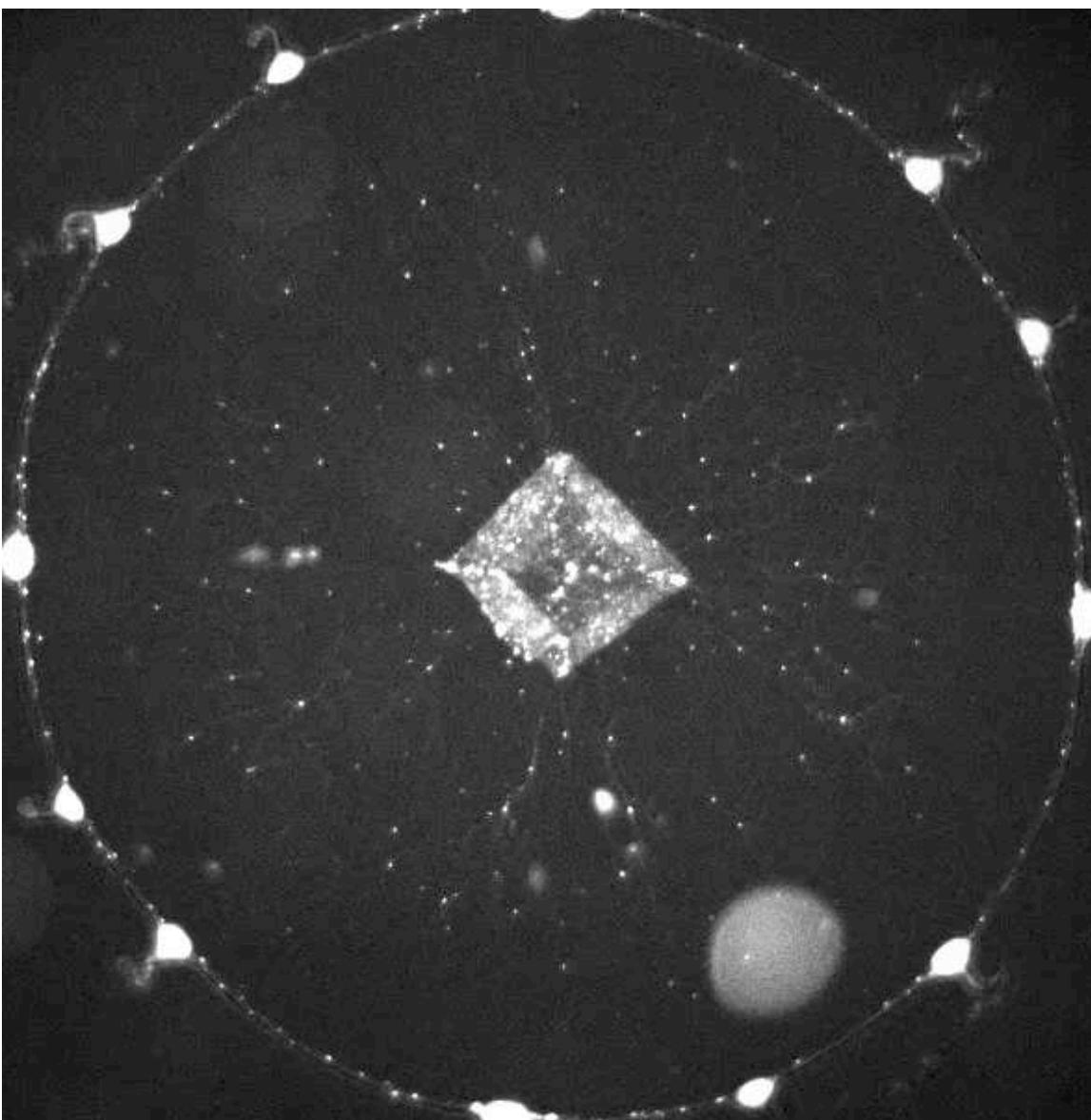
2. Topological defects & computation in liquid crystals

- nematic bits
- fractional defects and ‘anyonic’ braiding



How do living and soft matter systems compute ?

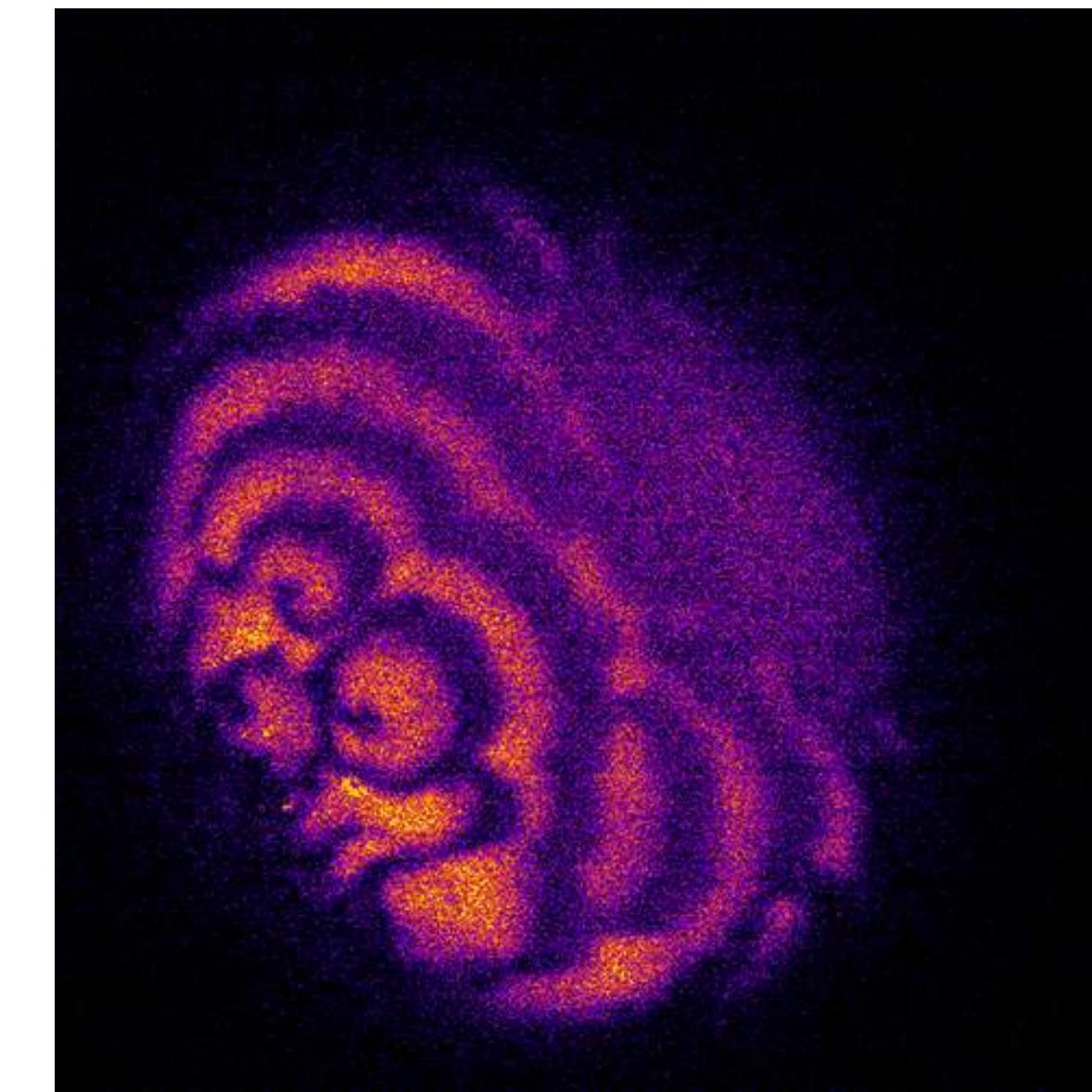
Jellyfish



Weissbourd lab (MIT Biology)

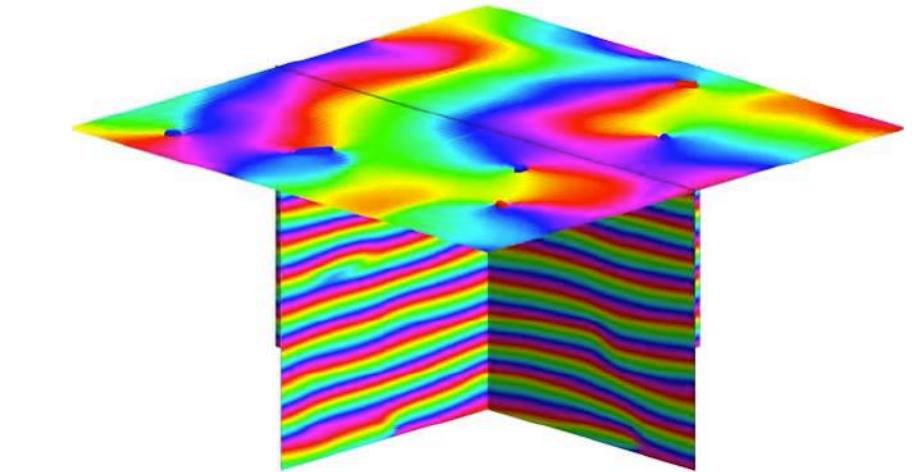
In progress

Starfish oocyte



Fakhri lab (MIT Physics)

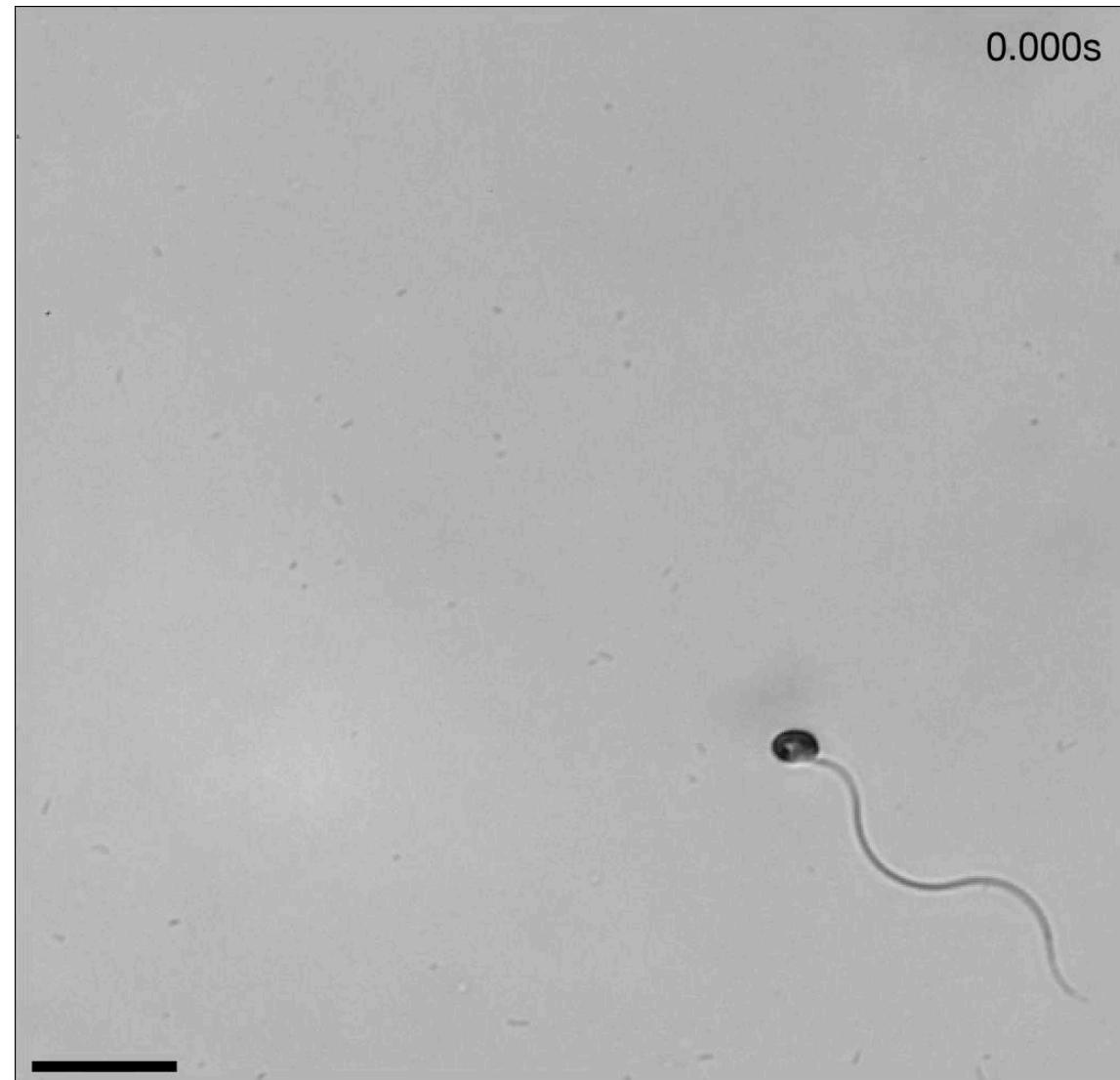
Nature Physics 2020
PNAS 2021



Part 1:

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

Pterosperma alga



Wang Lab (Exeter Uni)

μm

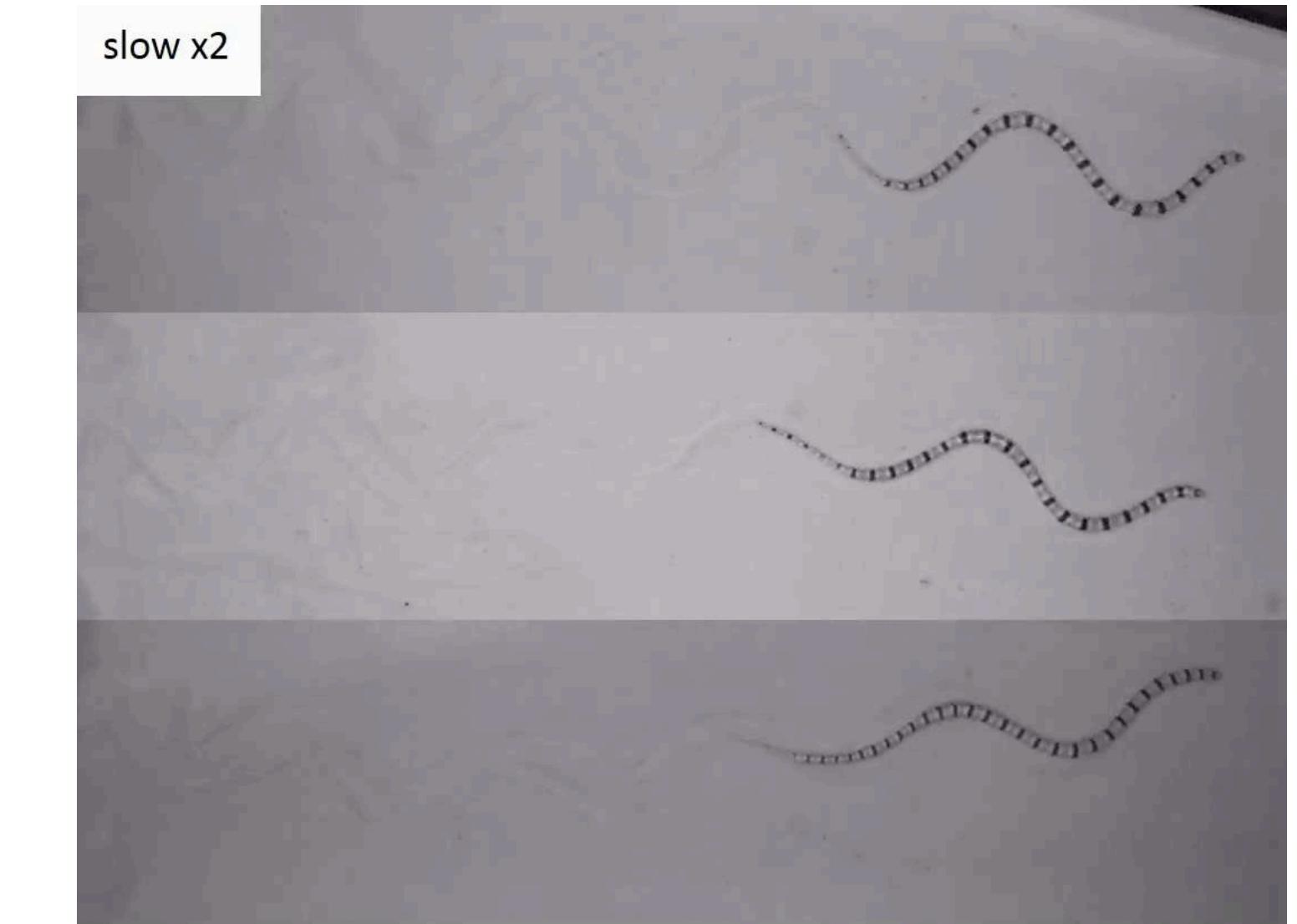
C elegans worm



Flavell Lab (MIT BCS)

mm

Desert snake



Goldman Lab (Georgia Tech)

m

Would like to ...

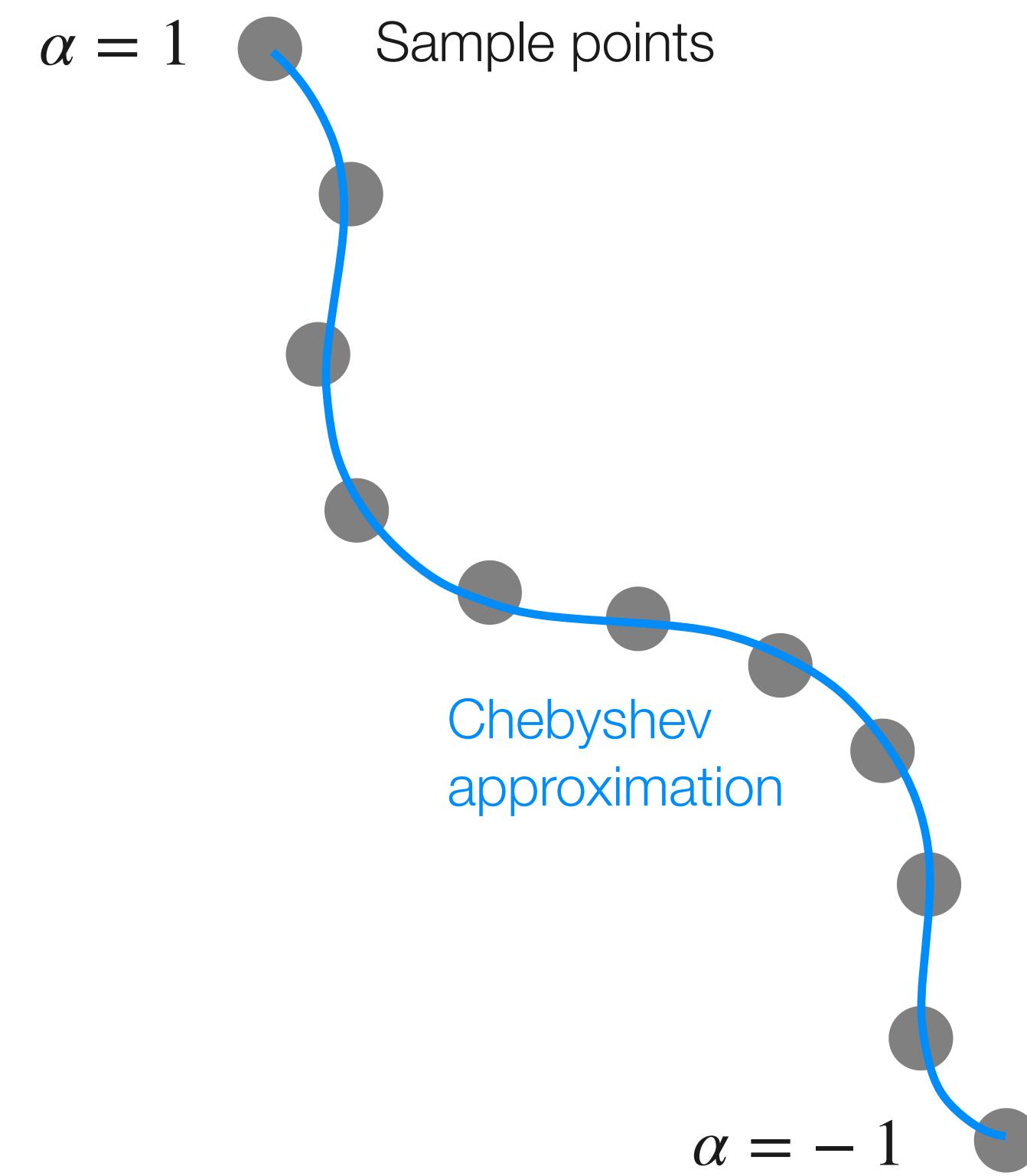
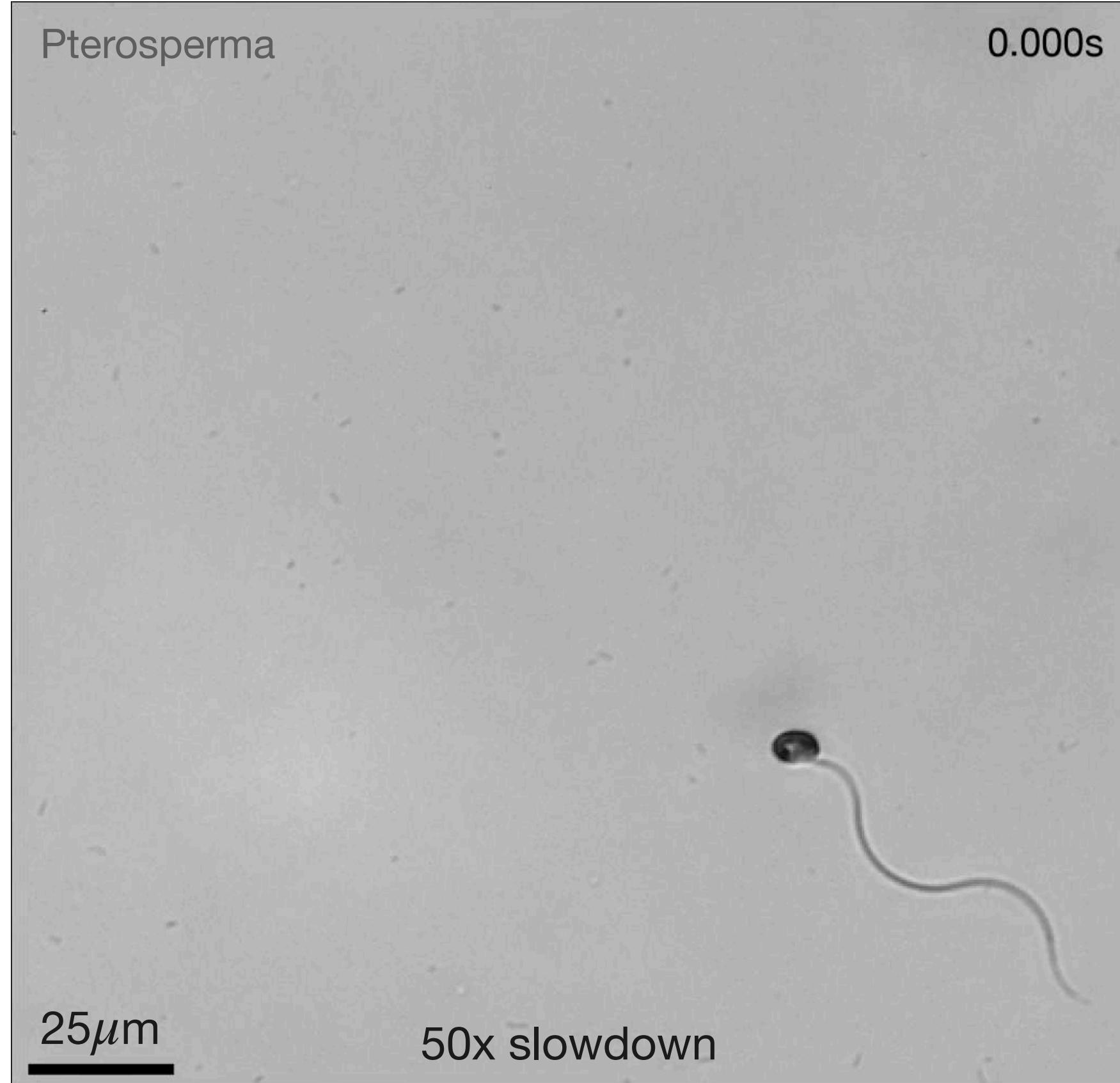
- Identify suitable collective ('slow') variables (low-dimensional representations)
- Identify and characterize different biological / biophysical states
- Infer dynamical models directly from video data
- Link biological signaling to physical dynamics



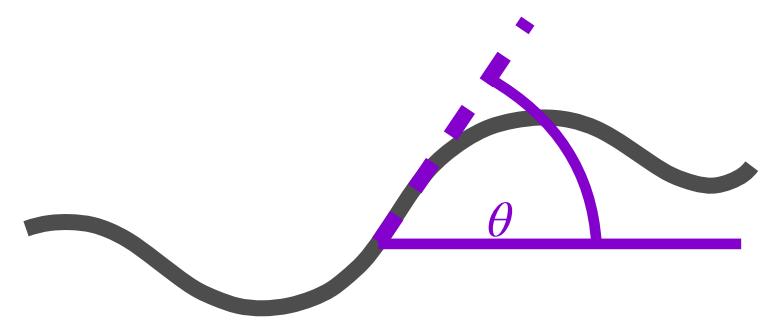
Spectral representation of biological dynamics & behaviors



Kirsty Wan Alex Boggon

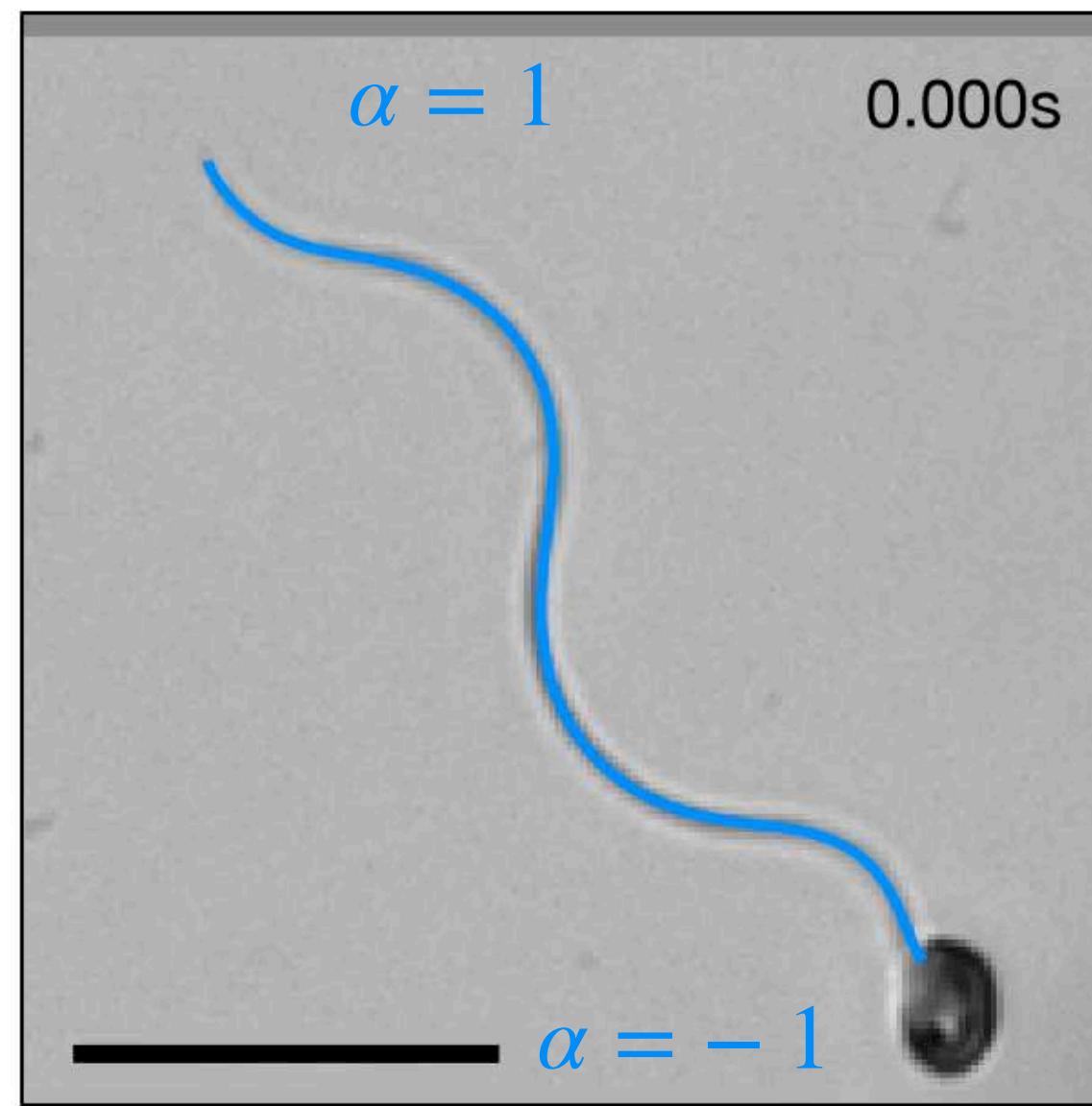


Spectral representation of biological dynamics & behaviors

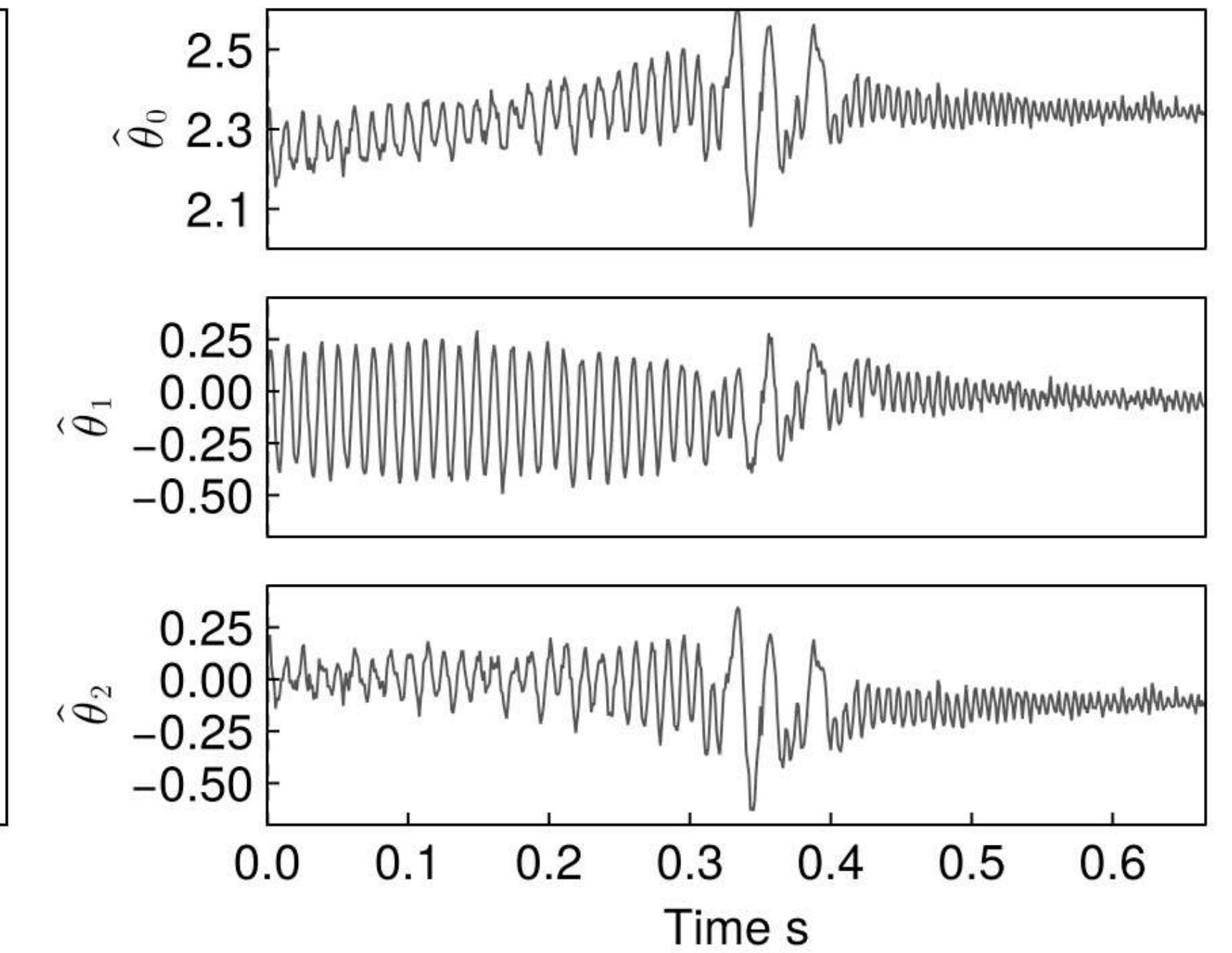
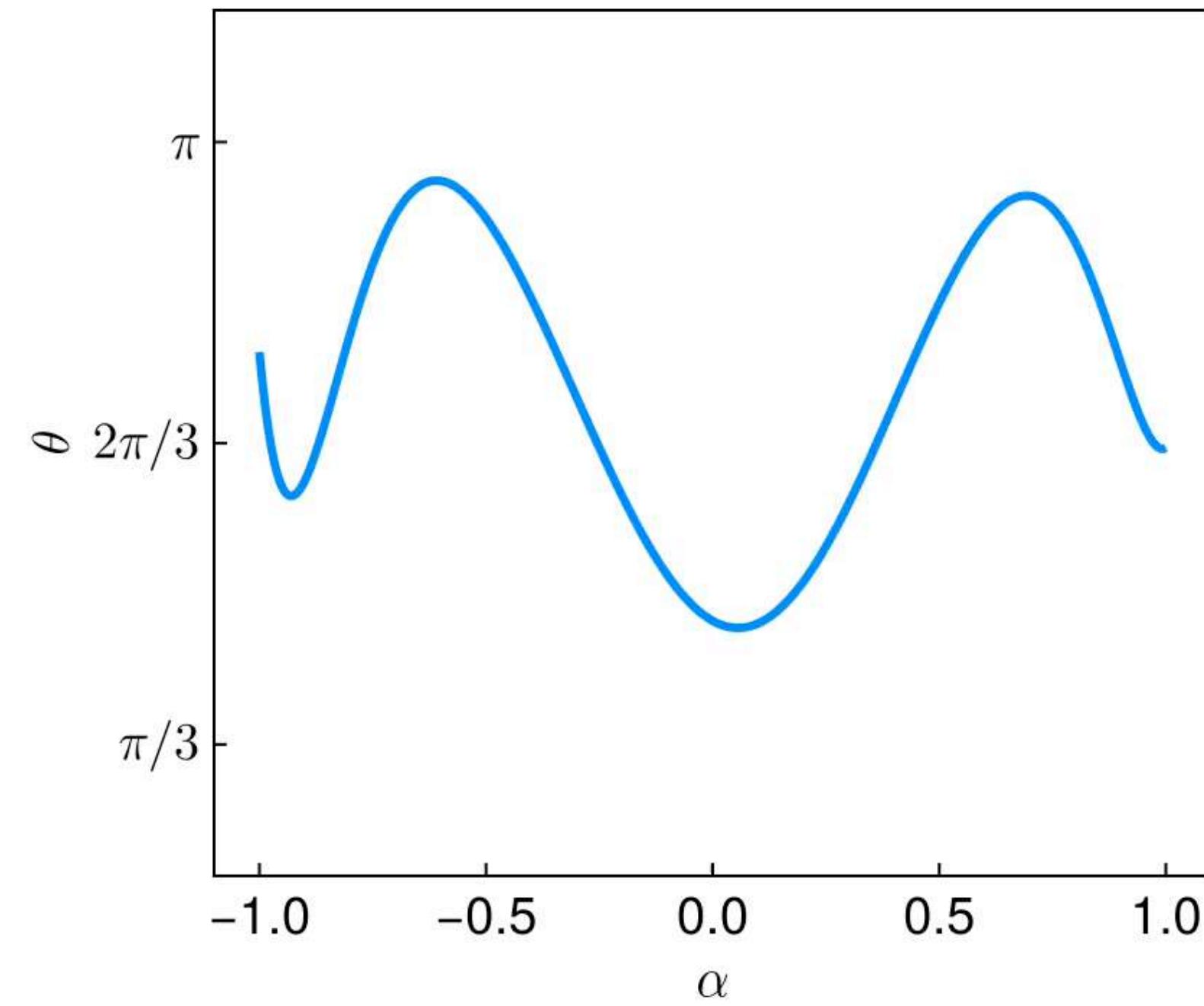


$$\theta(t, \alpha) \approx \sum_{n=1}^N \hat{\theta}_n(t) T_n(\alpha) = \hat{\theta}_1(t) + \hat{\theta}_2(t) + \hat{\theta}_3(t) + \hat{\theta}_4(t) + \dots$$

Orthogonal polynomial basis



Sparse 10-mode representation

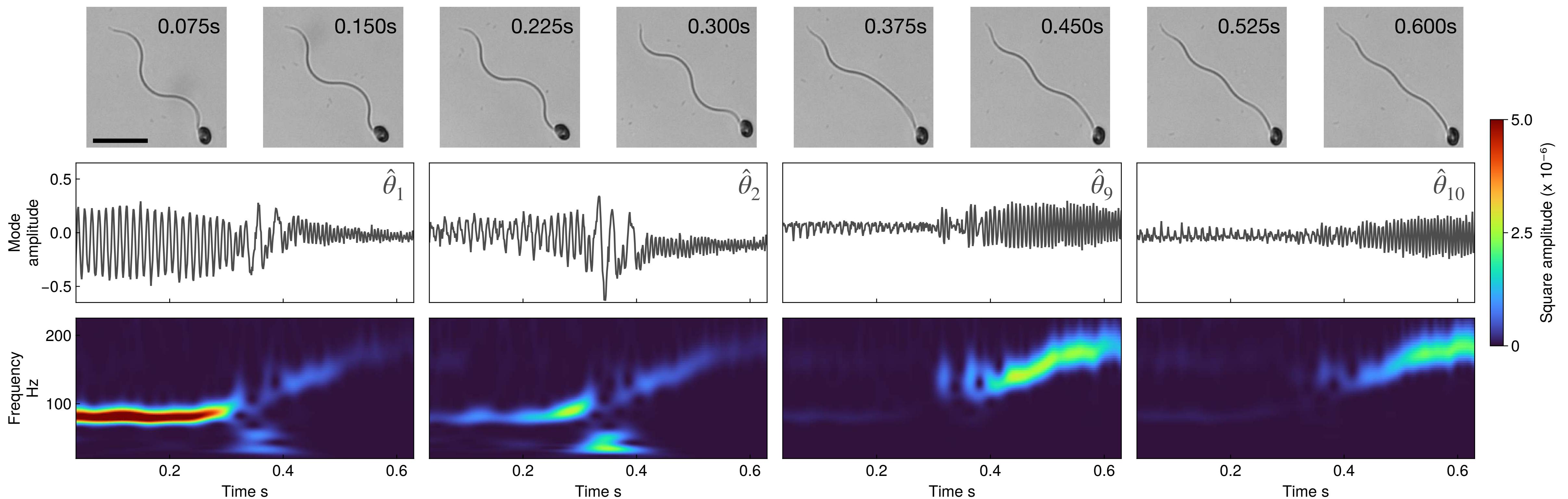
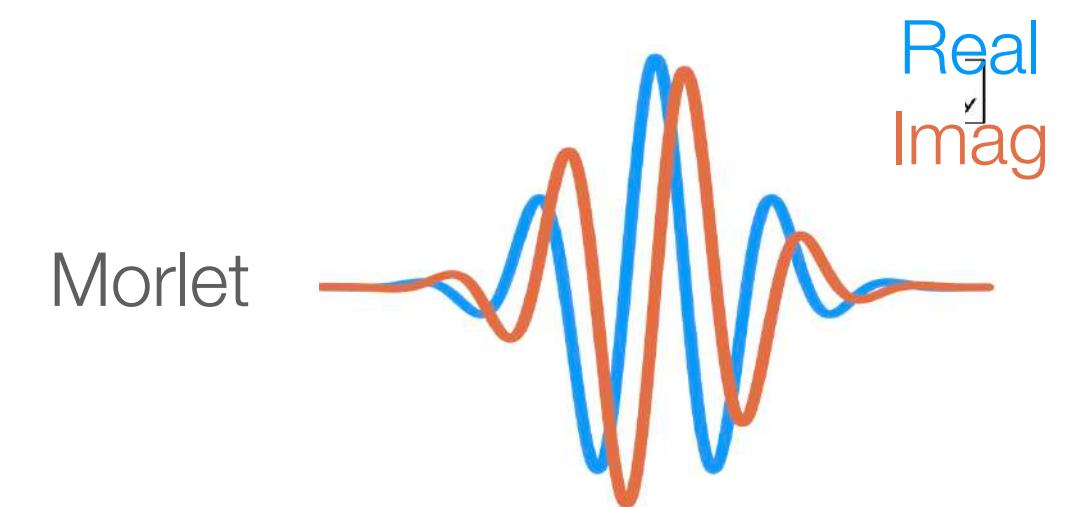


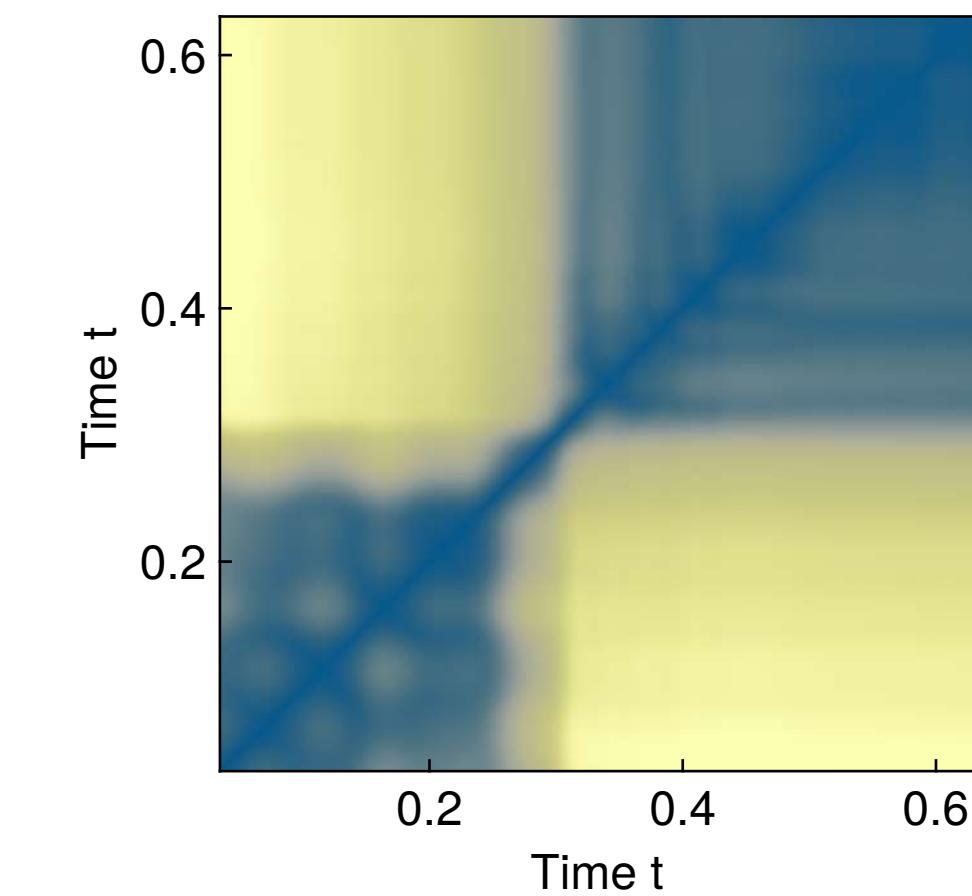
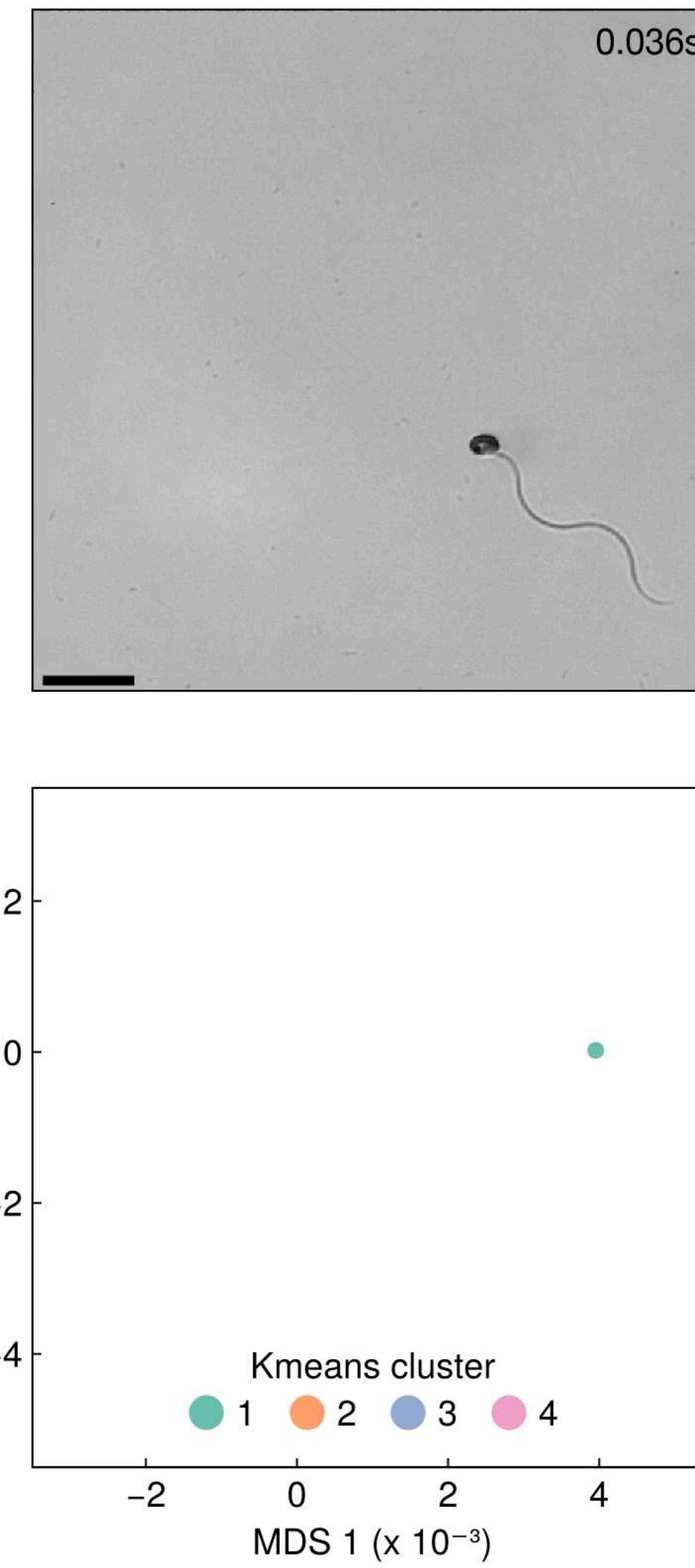
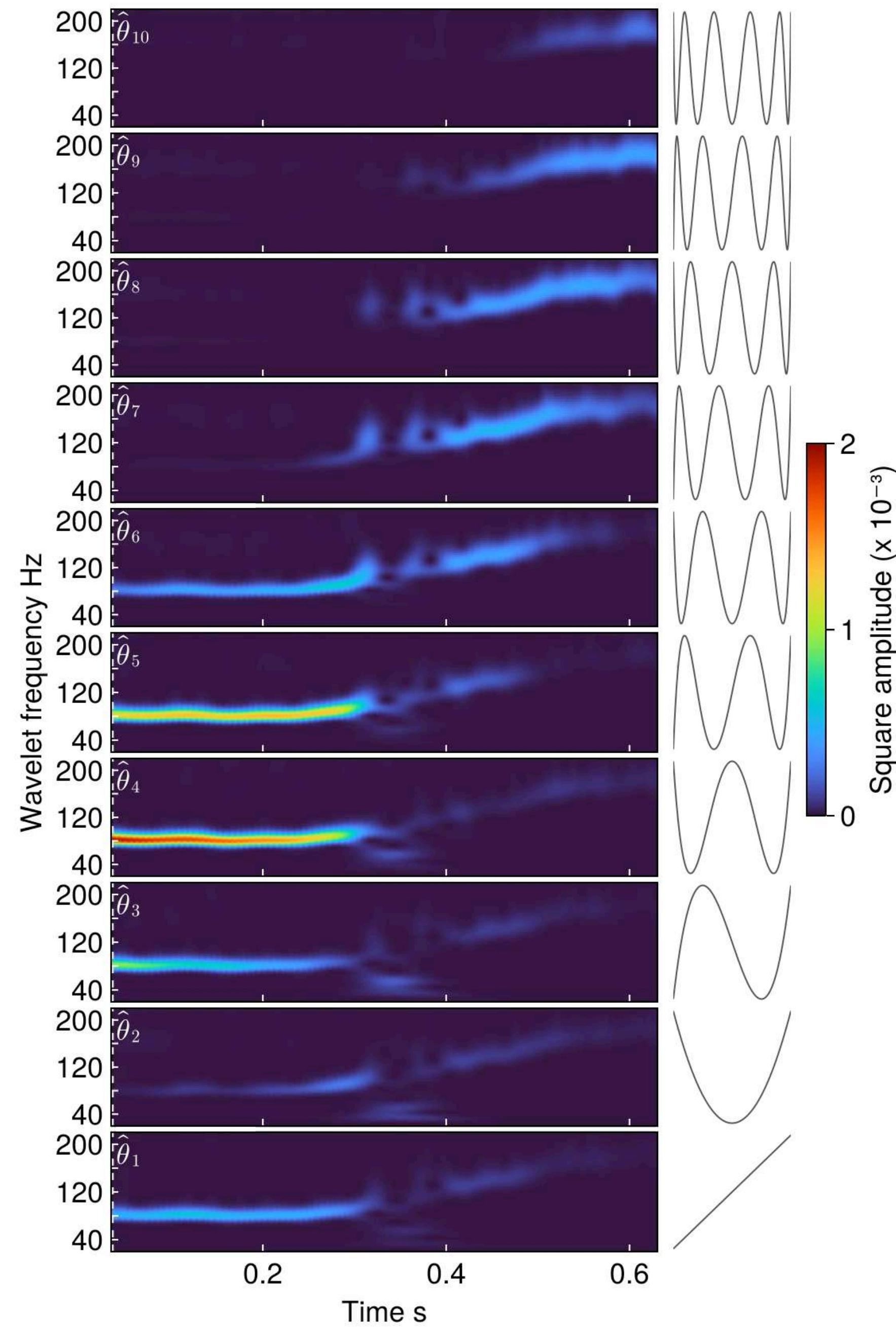
Wavelet analysis in time

Convolve signal with shifts and scales of wavelet

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

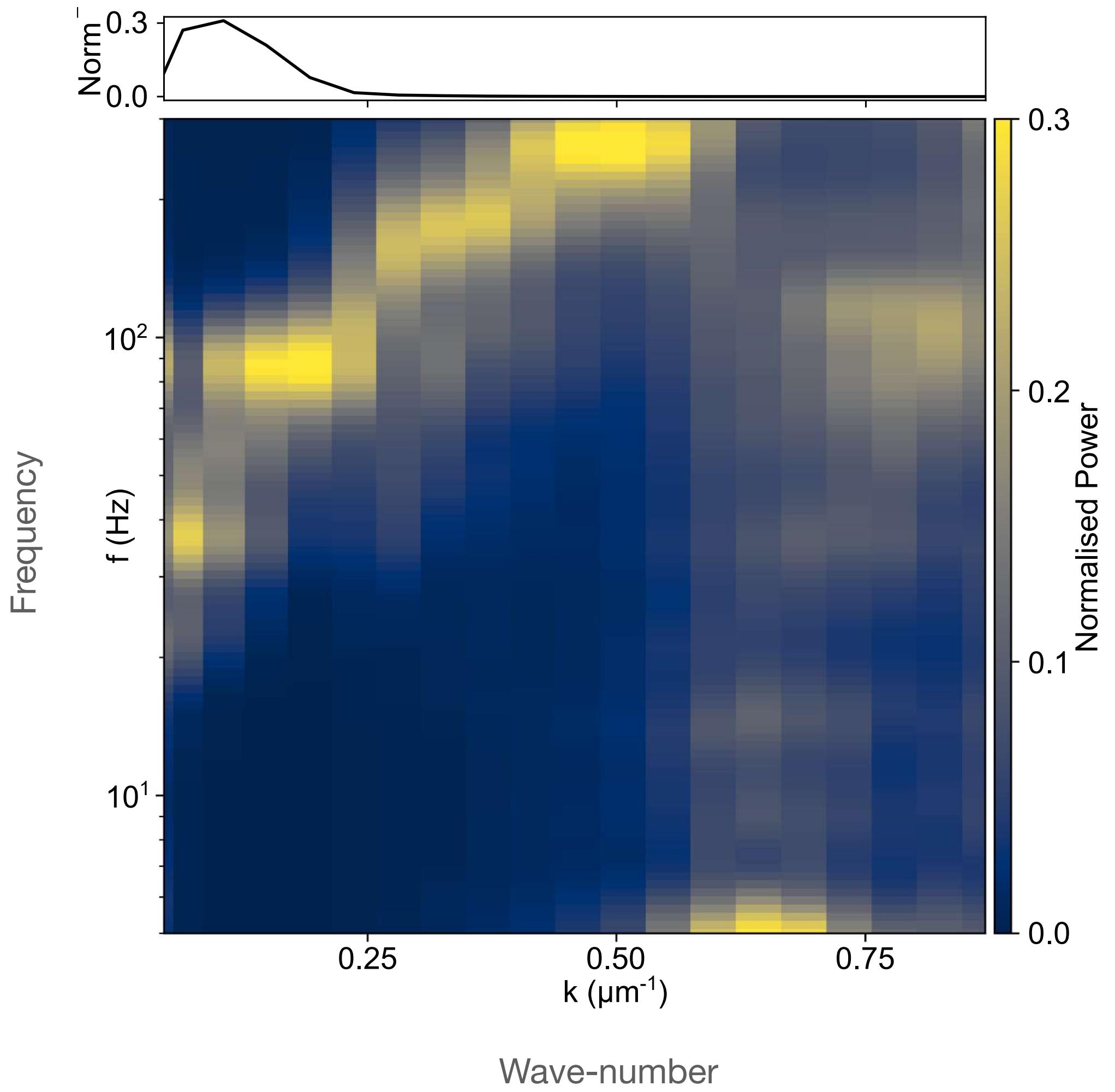
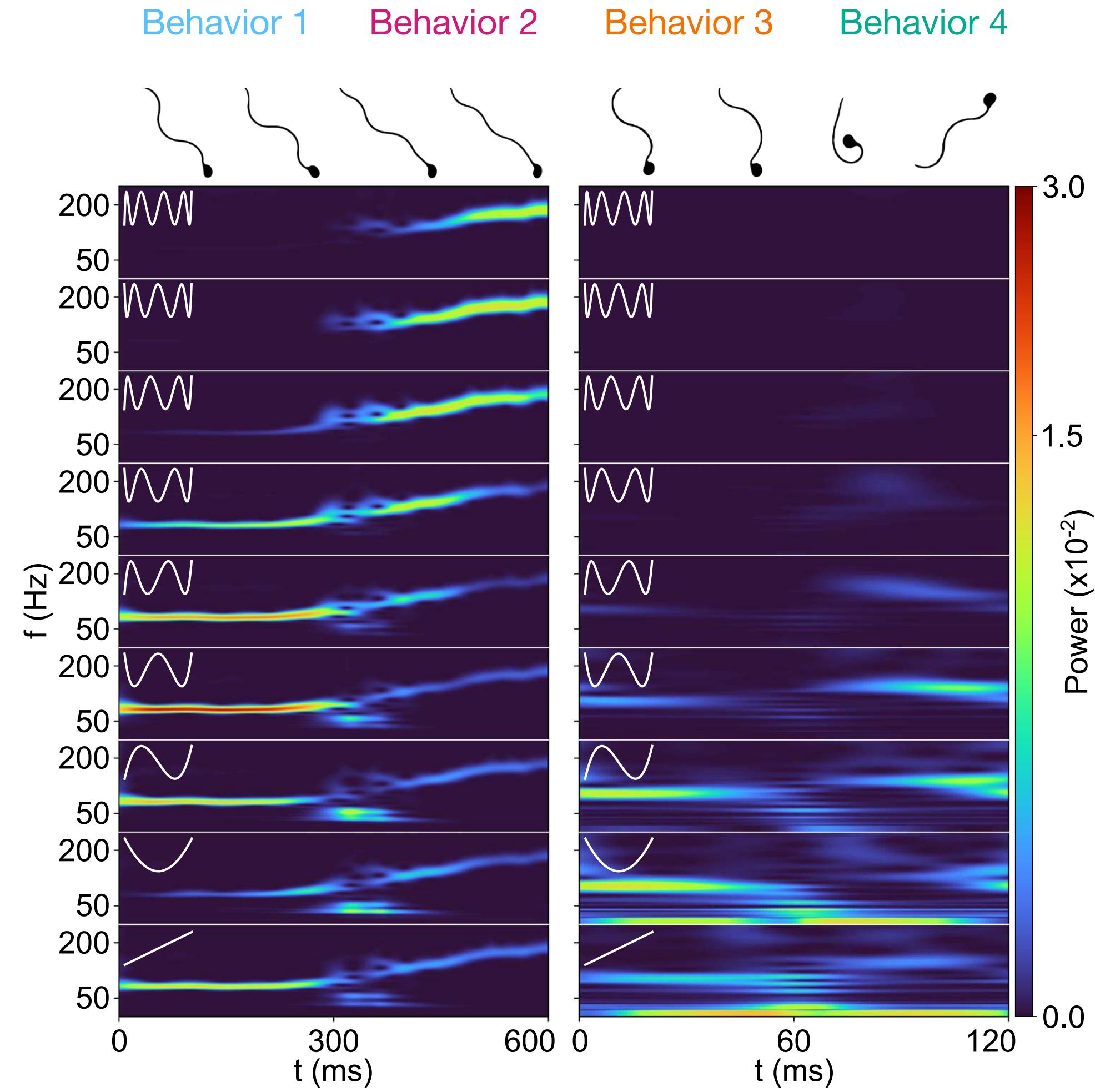
temporal location
wavelet frequency





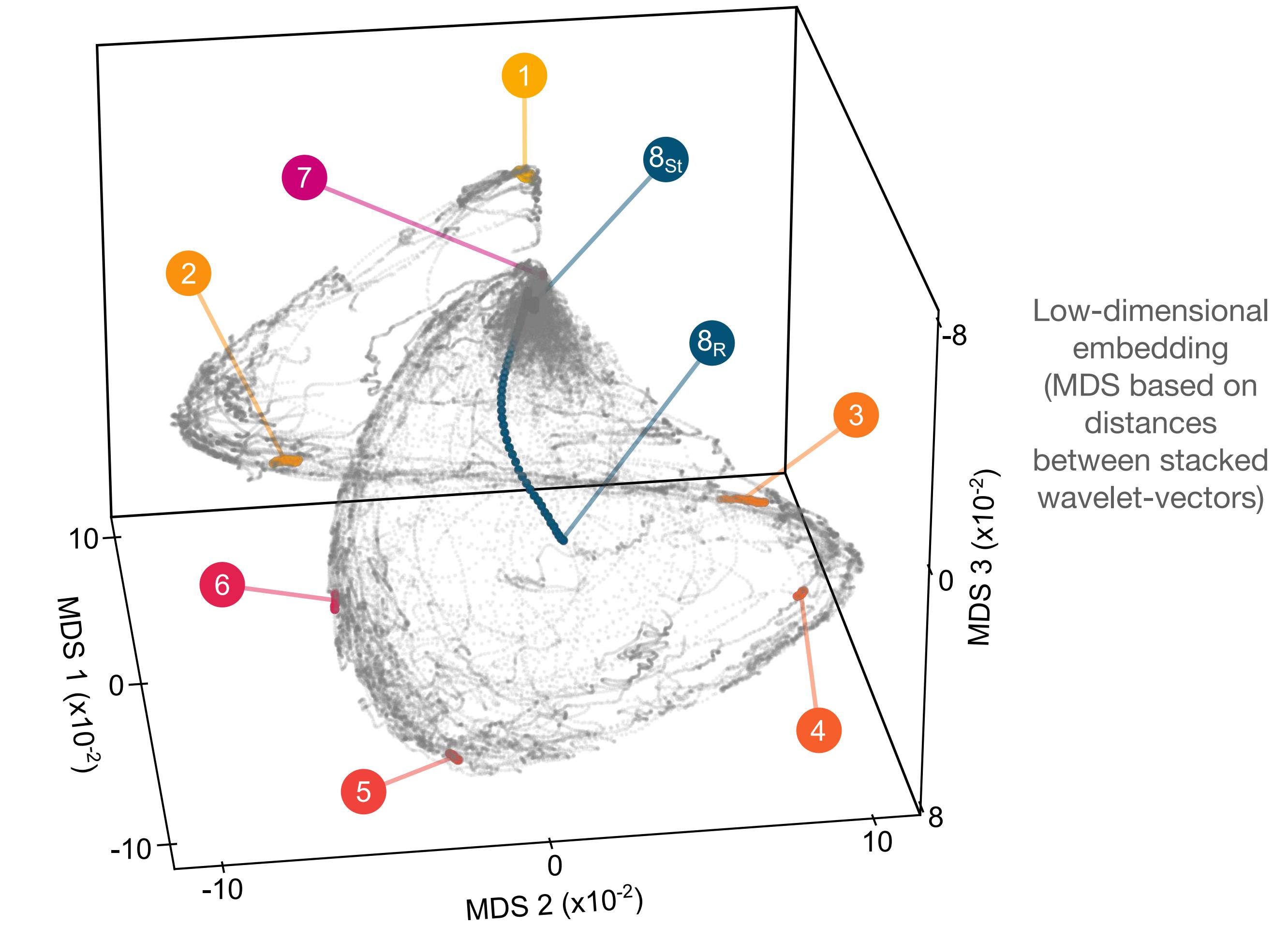
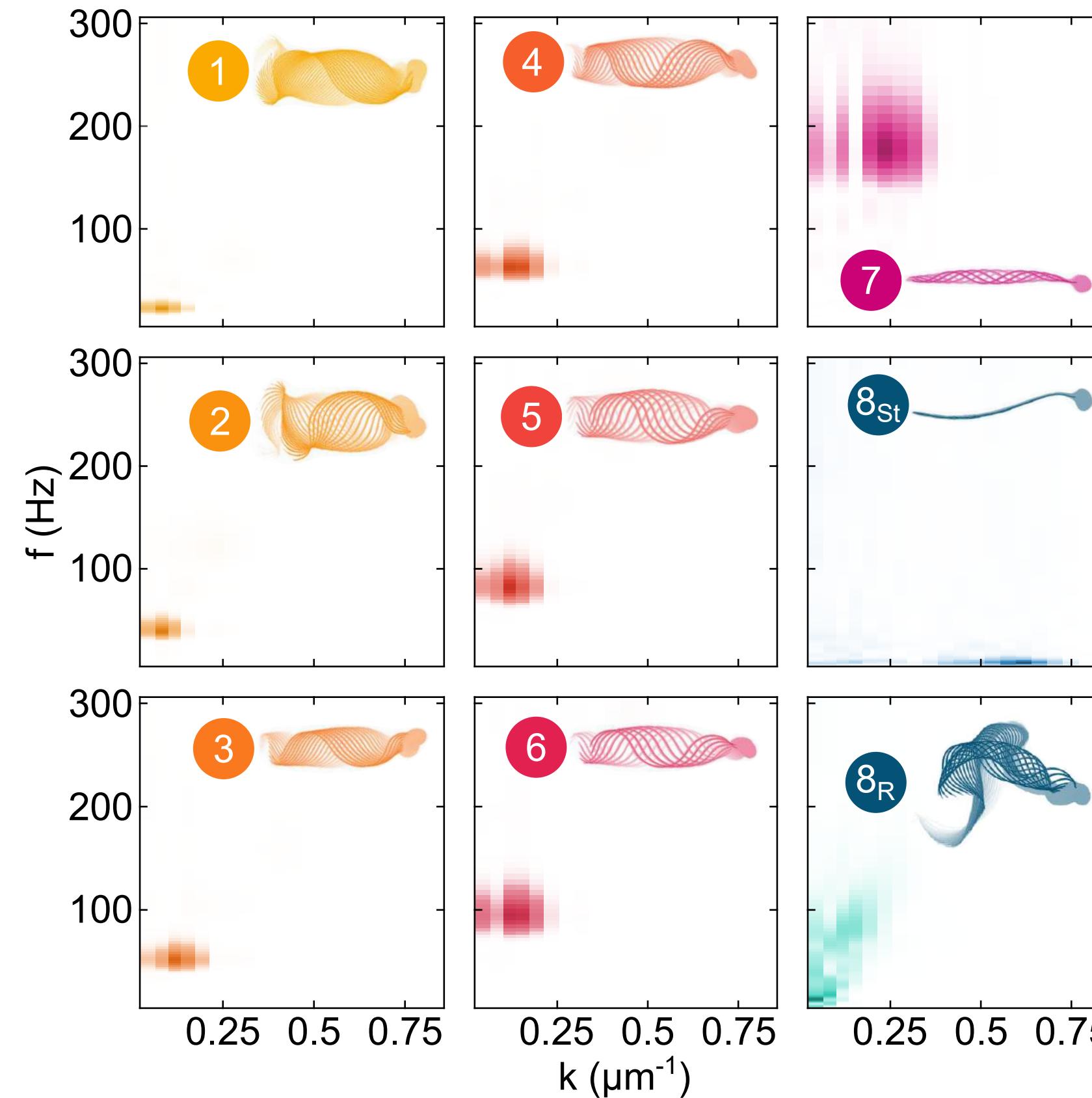
Instantaneous wavelet state encodes instantaneous dynamical behavior

Dispersion relation



Behavioral manifold

Wave-modes
activated
in a specific
behavioral state



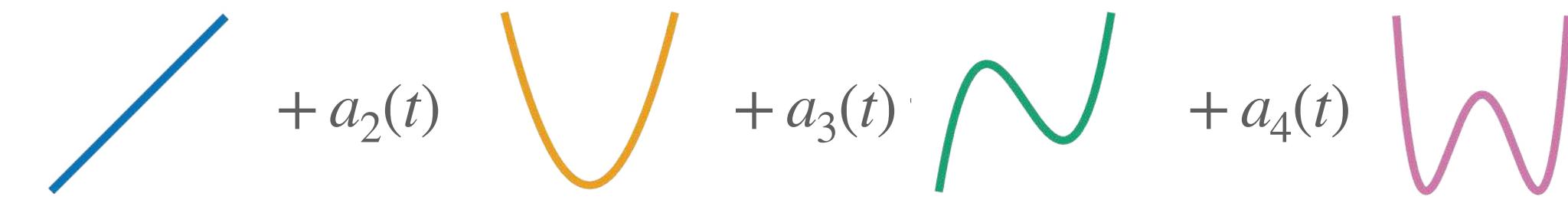
Generalizes across systems & scales

Roundworm

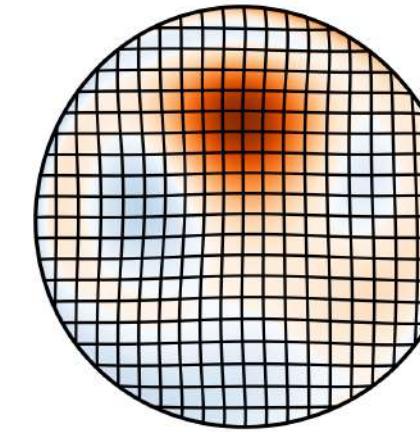
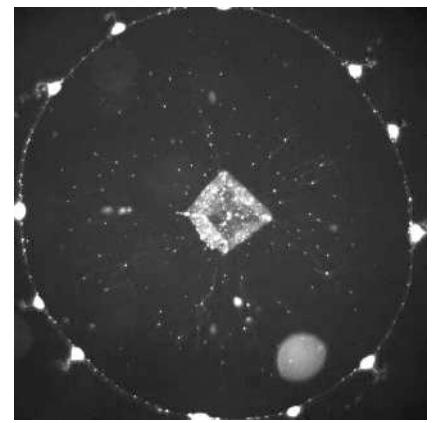


$$= a_0(t) \text{ — } + a_1(t)$$

Legendre / Chebyshev / data-informed orthogonal polynomials

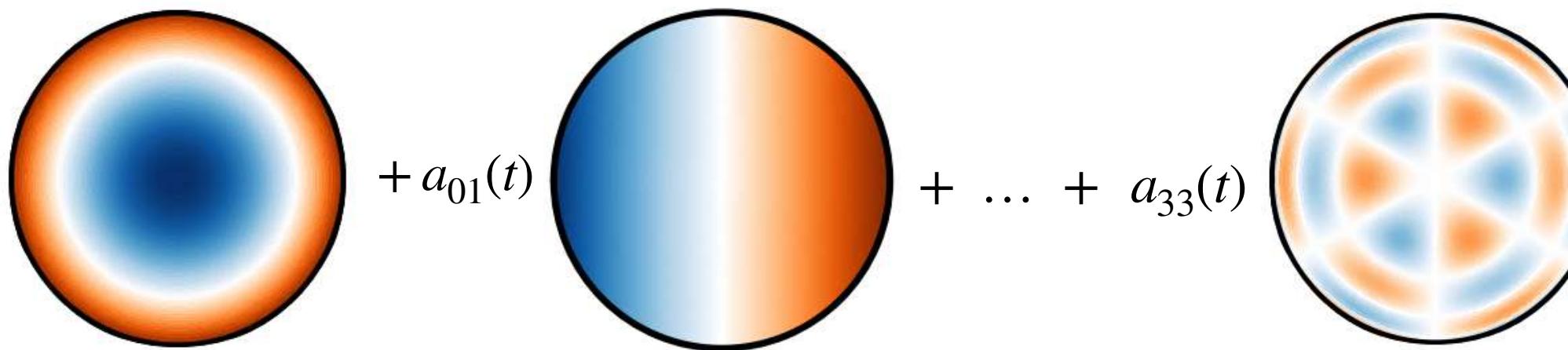


Jellyfish

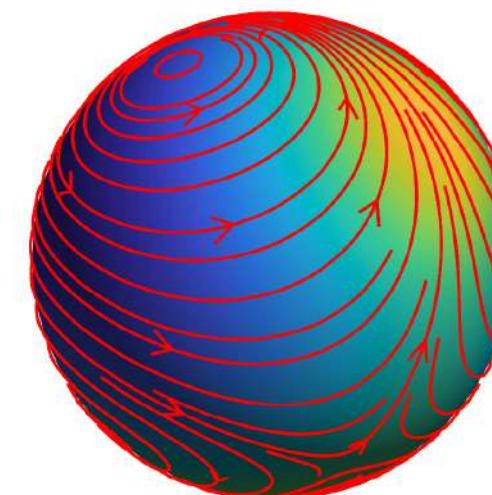
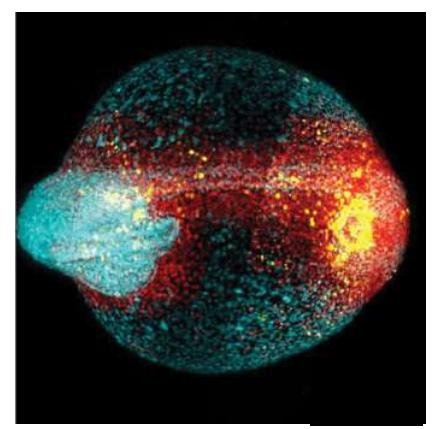


$$= a_{00}(t) + a_{10}(t)$$

Fourier-Jacobi basis functions

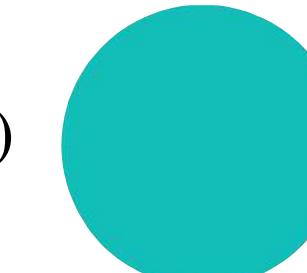


Zebrafish

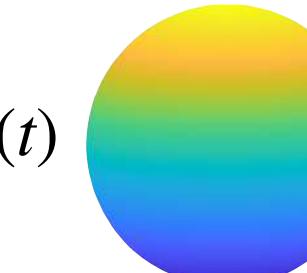


$$=$$

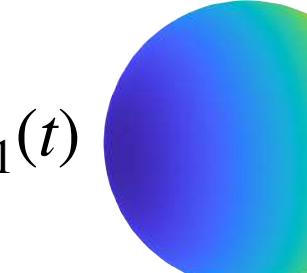
$$a_{00}(t)$$



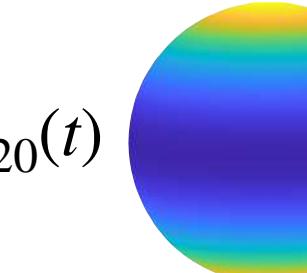
$$+ a_{10}(t)$$



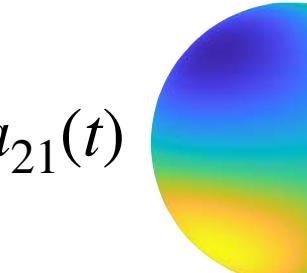
$$+ a_{11}(t)$$



$$+ a_{20}(t)$$



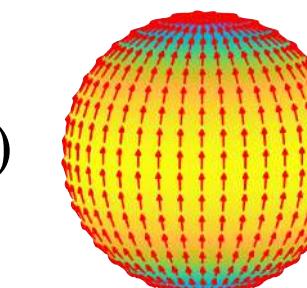
$$+ a_{21}(t)$$



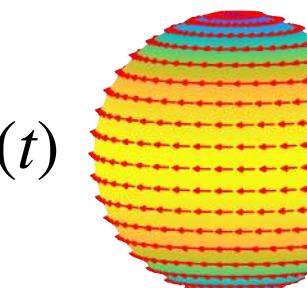
$$+ \dots$$

Scalar spherical harmonics

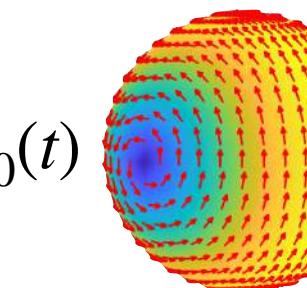
$$b_{10}(t)$$



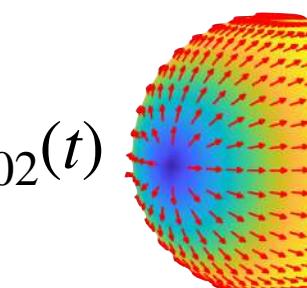
$$+ b_{01}(t)$$



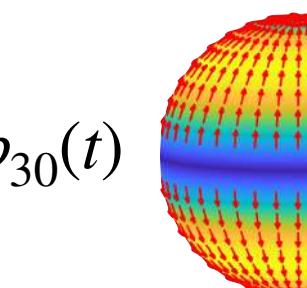
$$+ b_{20}(t)$$



$$+ b_{02}(t)$$



$$+ b_{30}(t)$$

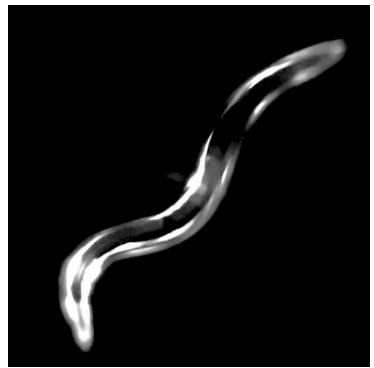


$$+ \dots$$

Vector spherical harmonics

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

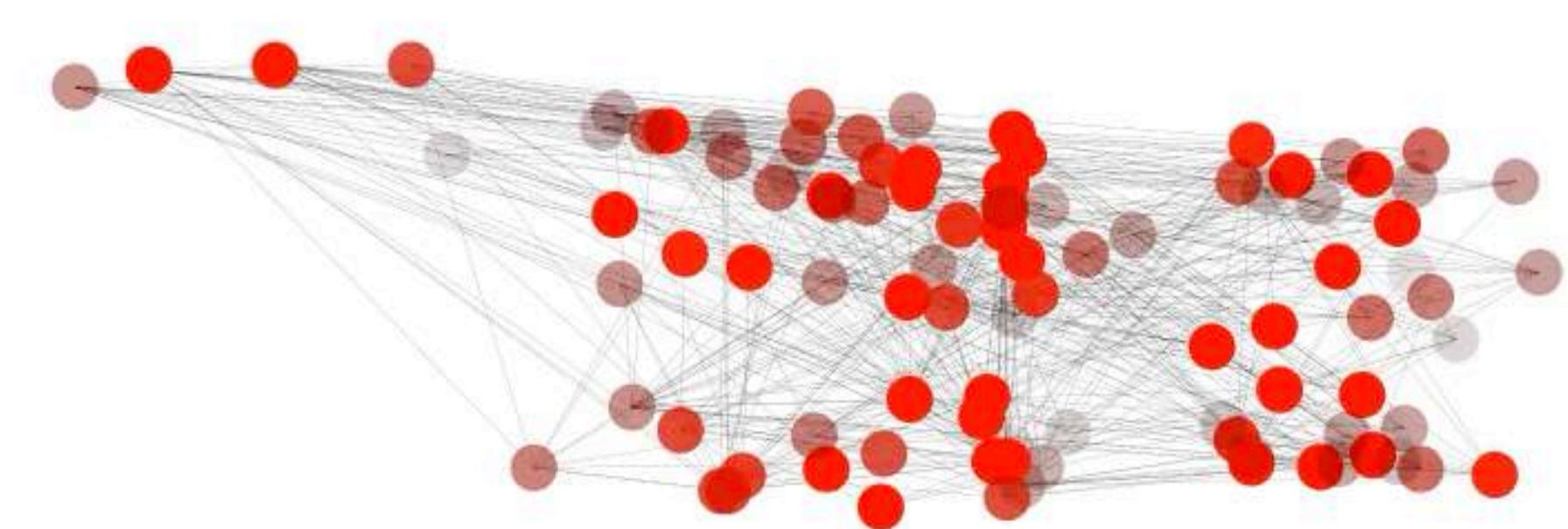
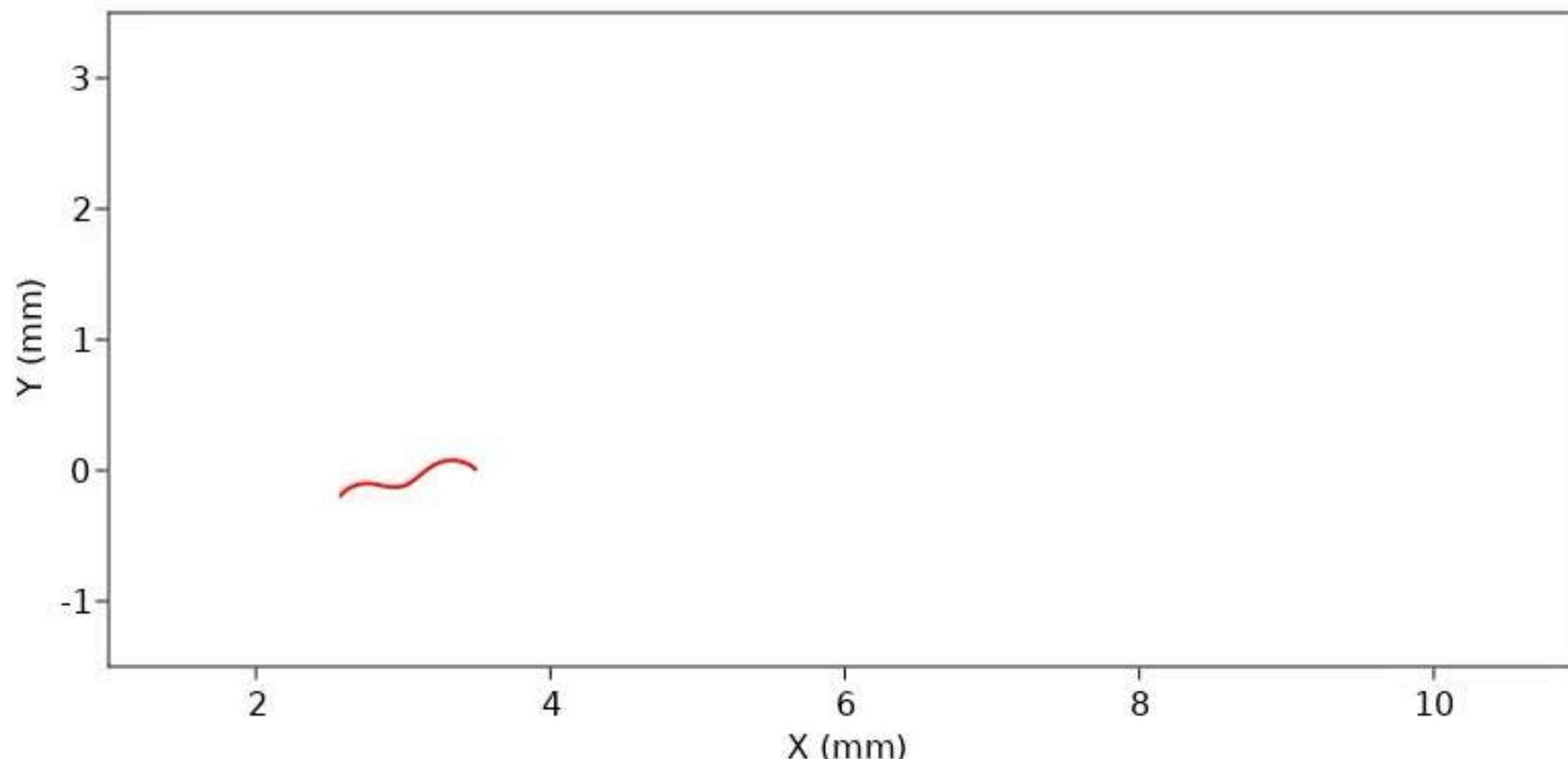
Experimental motivation



C elegans

physical locomotion dynamics \Leftrightarrow 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, ...)

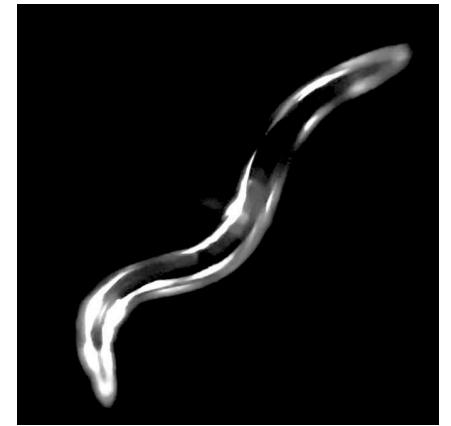
Stochastic

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

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

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

maximum likelihood estimation of p

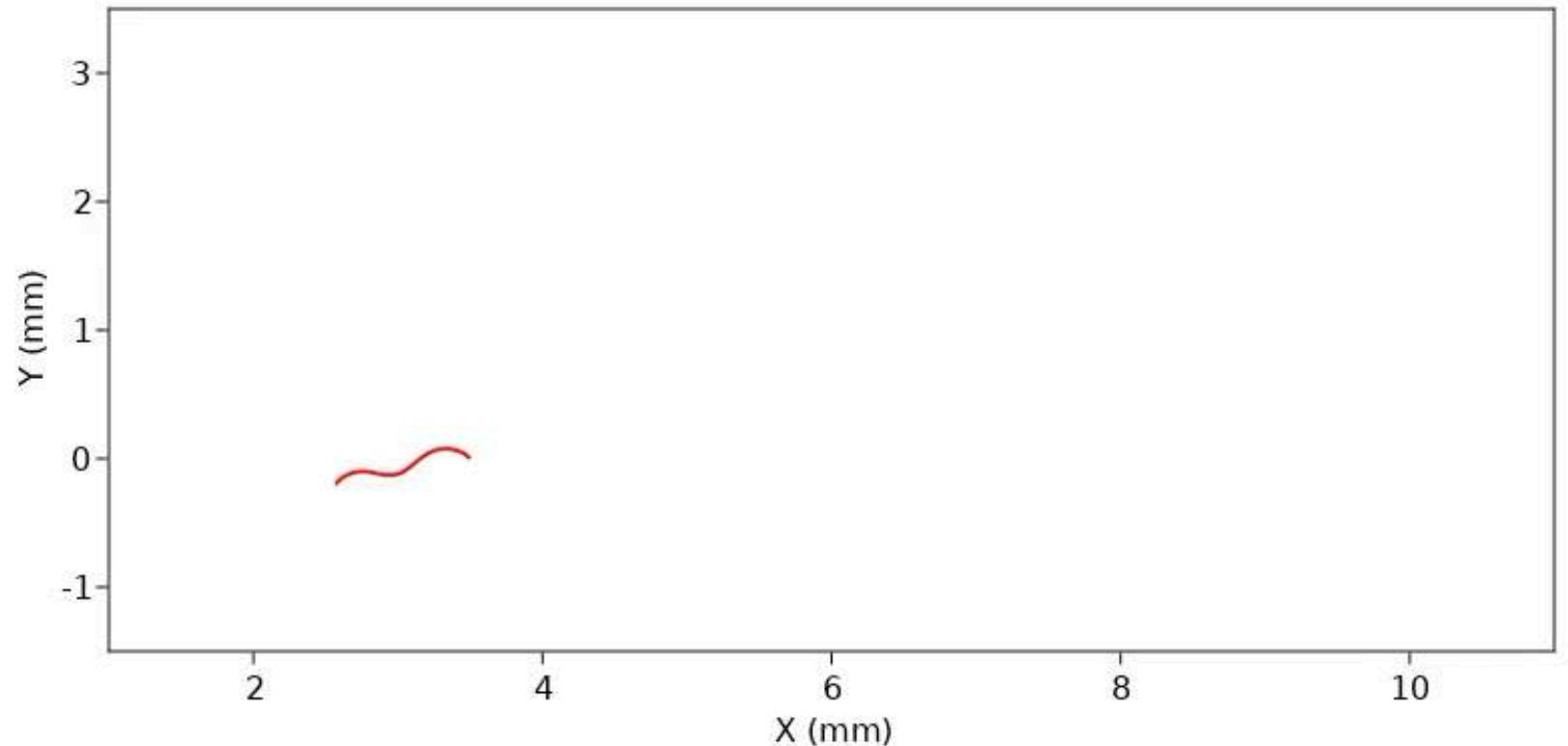


Learning linear models for undulatory locomotion dynamics



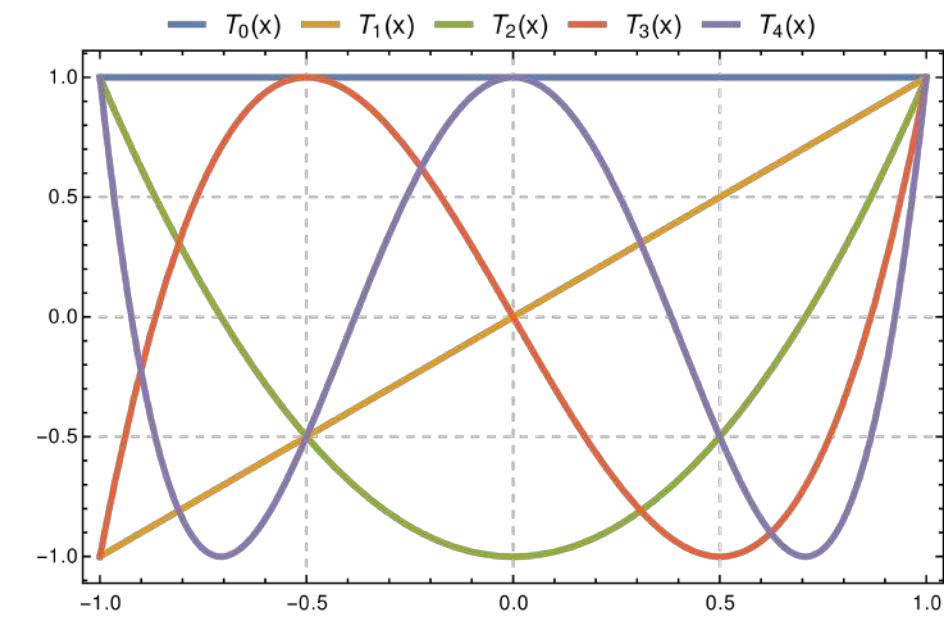
Flavell lab
(MIT BCS)

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

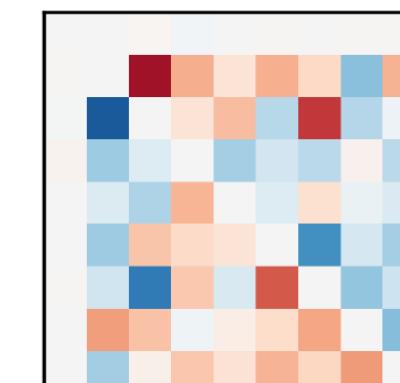
$$\frac{d}{dt} \begin{bmatrix} \hat{\mathbf{x}} \\ \hat{\mathbf{y}} \end{bmatrix} = \begin{bmatrix} \mathbf{M}^{(2)} \mathbf{W} & 0 \\ 0 & \mathbf{M}^{(2)} \mathbf{W} \end{bmatrix} \begin{bmatrix} \hat{\mathbf{x}} \\ \hat{\mathbf{y}} \end{bmatrix}$$

$$\frac{N \times (N - 1)}{2}$$

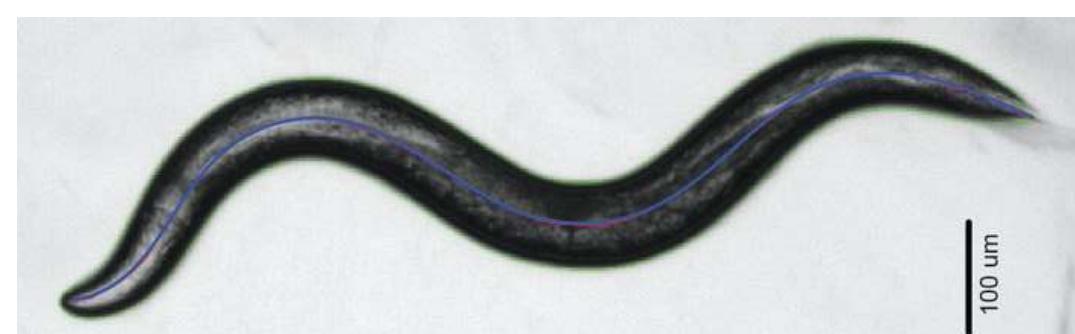


$$\mathbf{W}_{ij} = \int_{-1}^1 ds \left(\frac{dT_i}{ds} \frac{dT_j}{ds} \right)$$

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



Flavell lab (MIT Picower)



Planar curve

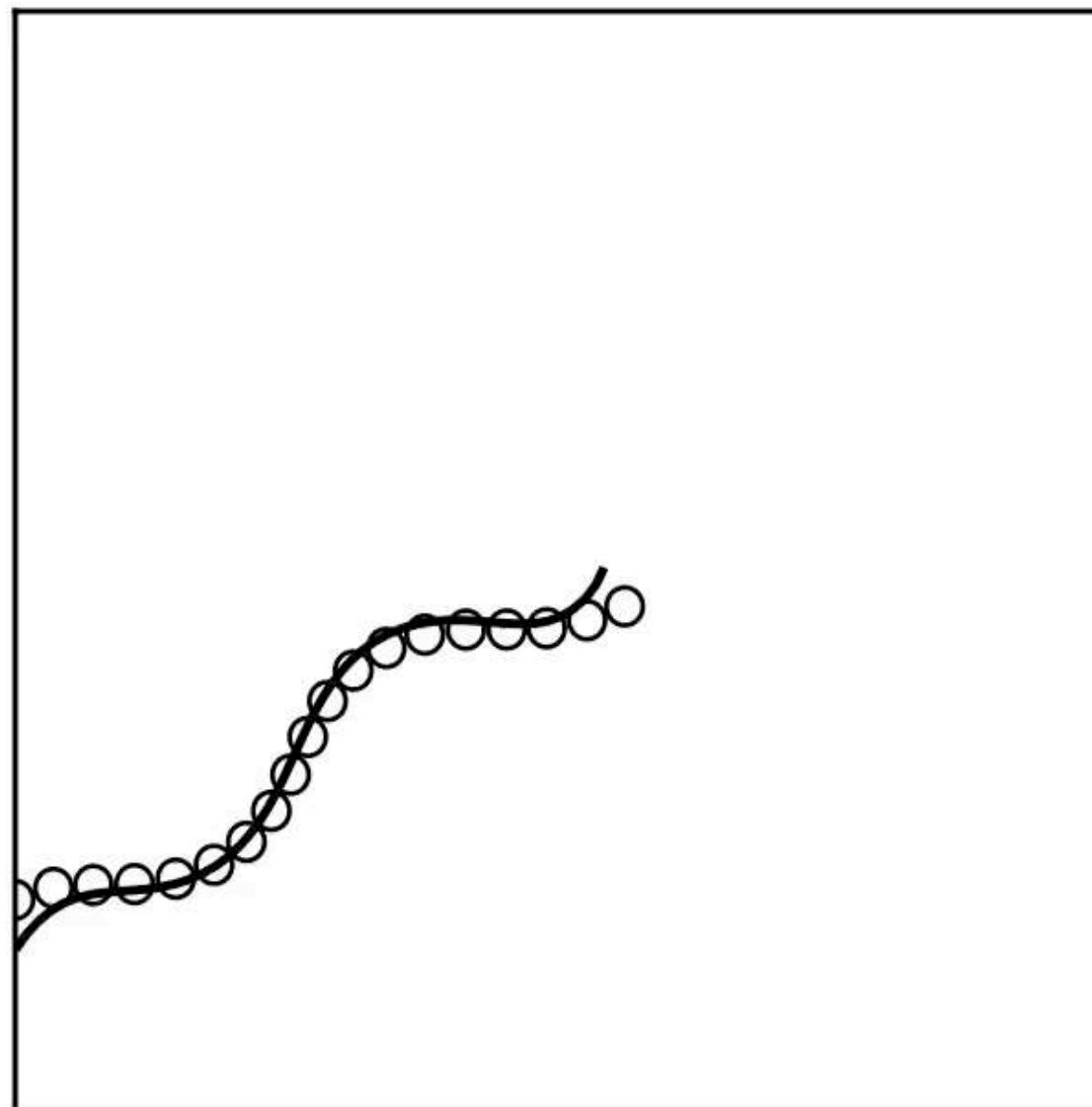
$$x(s, t) + iy(s, t) \quad \Rightarrow$$

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

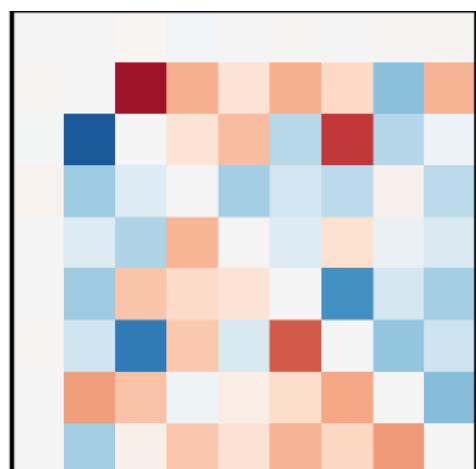
$$i\dot{\psi} = H\psi$$

Shape mode dynamics

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

Propulsion dynamics

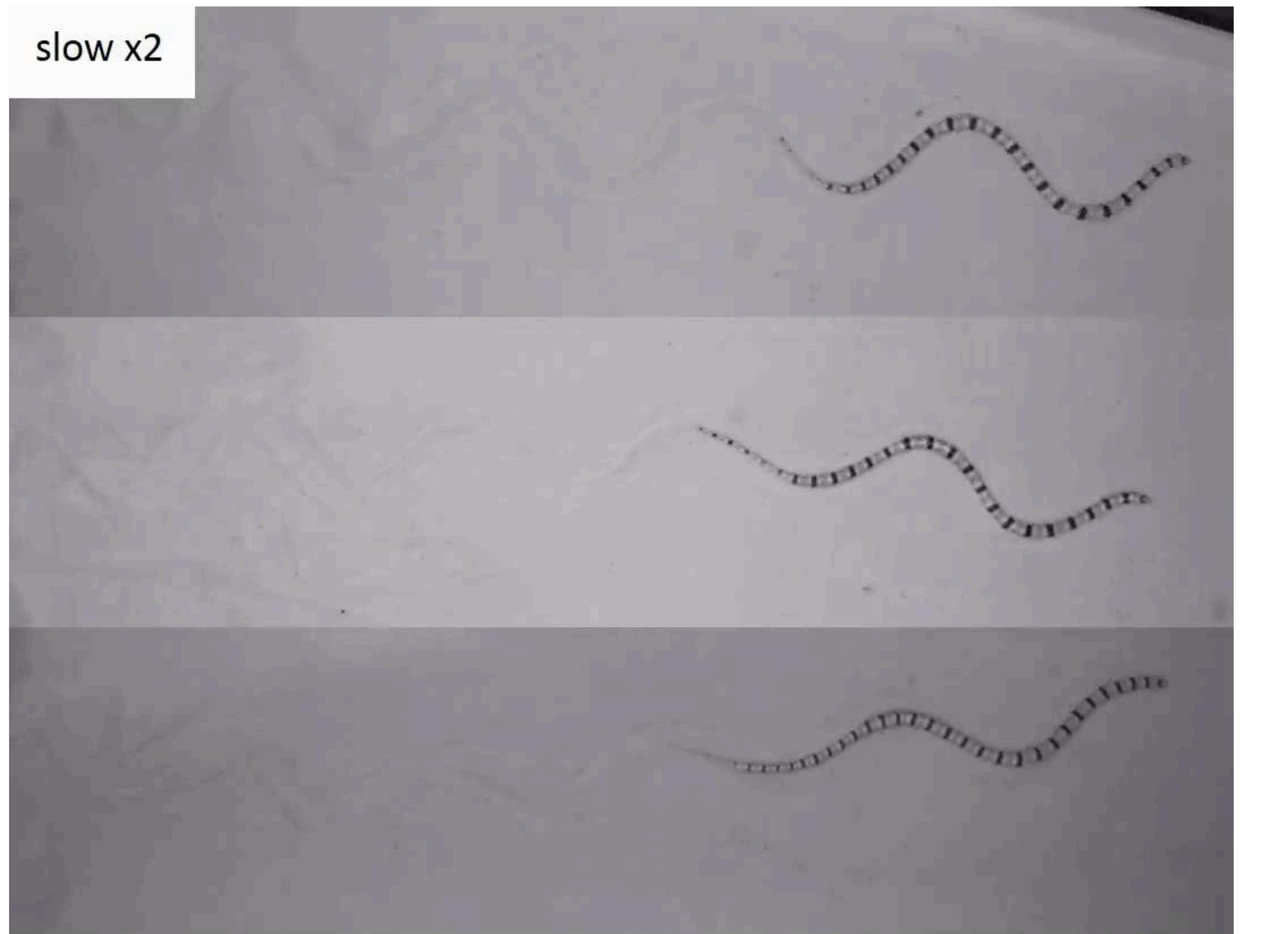
H



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

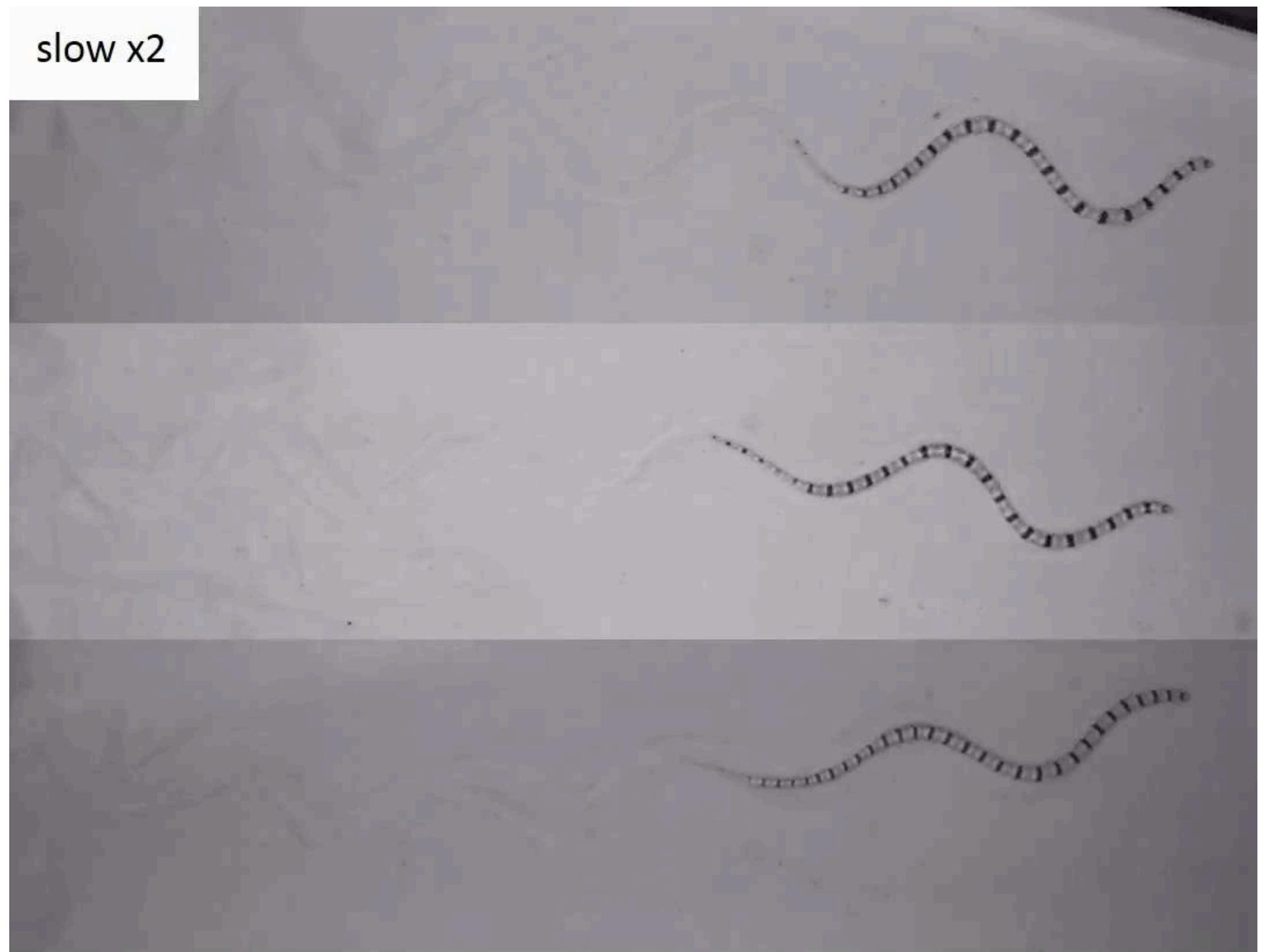
Straight motion

Desert snake



Goldman Lab (Georgia Tech), PNAS 2019

Desert snake



Toy snake

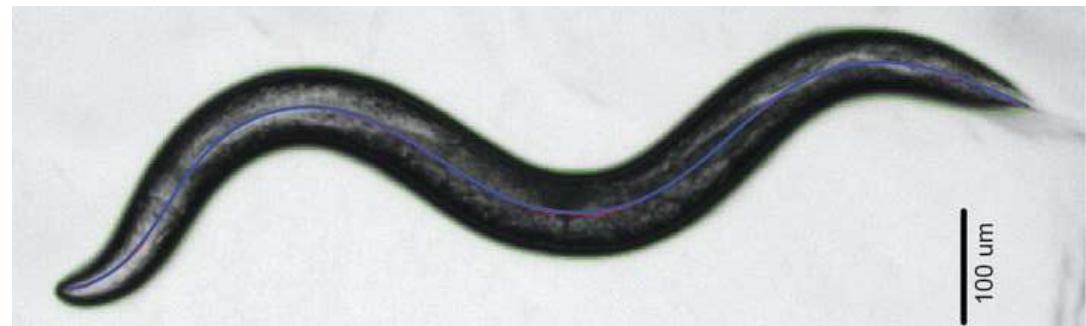


Goldman Lab (Georgia Tech), PNAS 2019

Cohen*, Hastewell* et al, PRL 2023



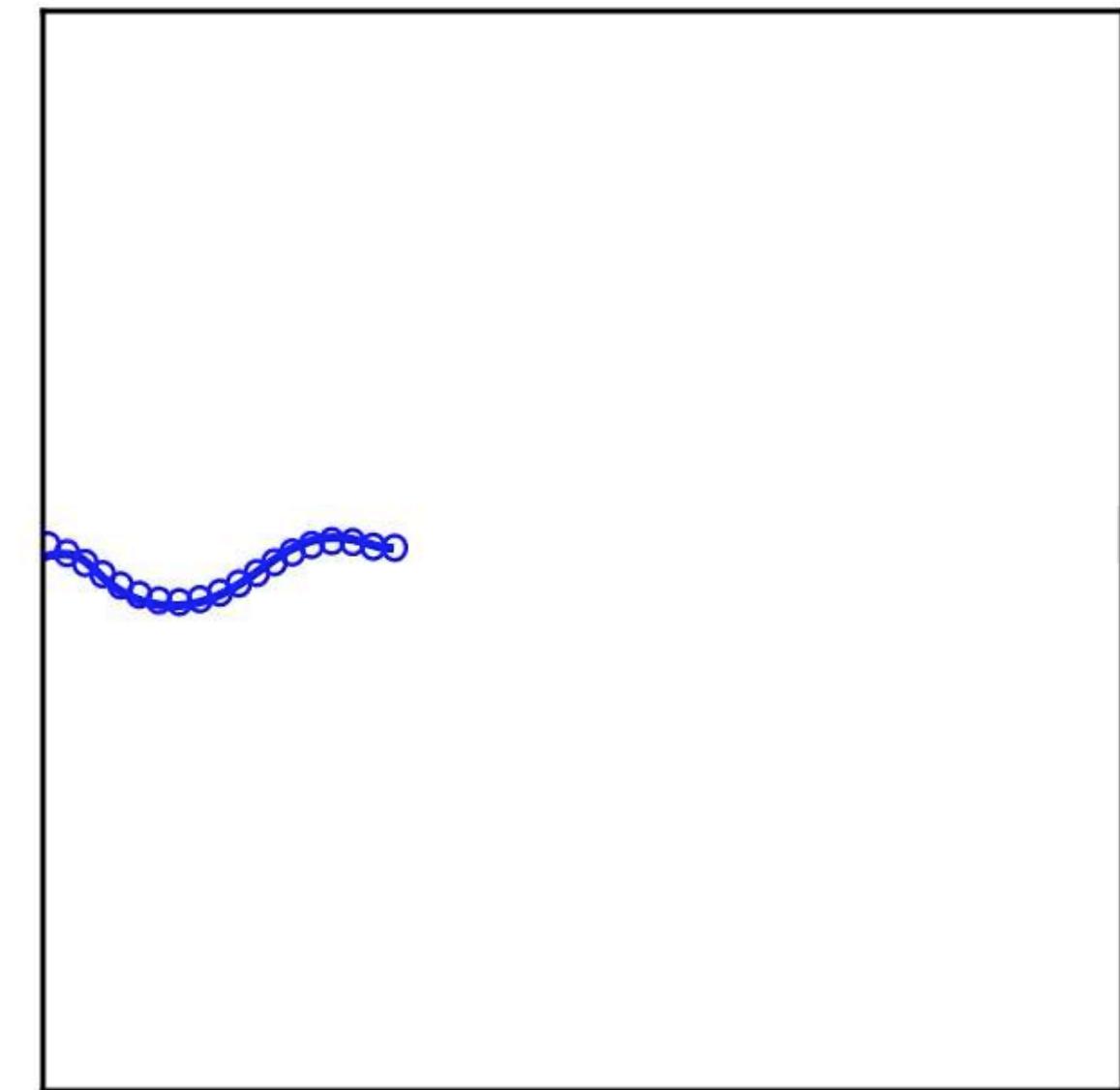
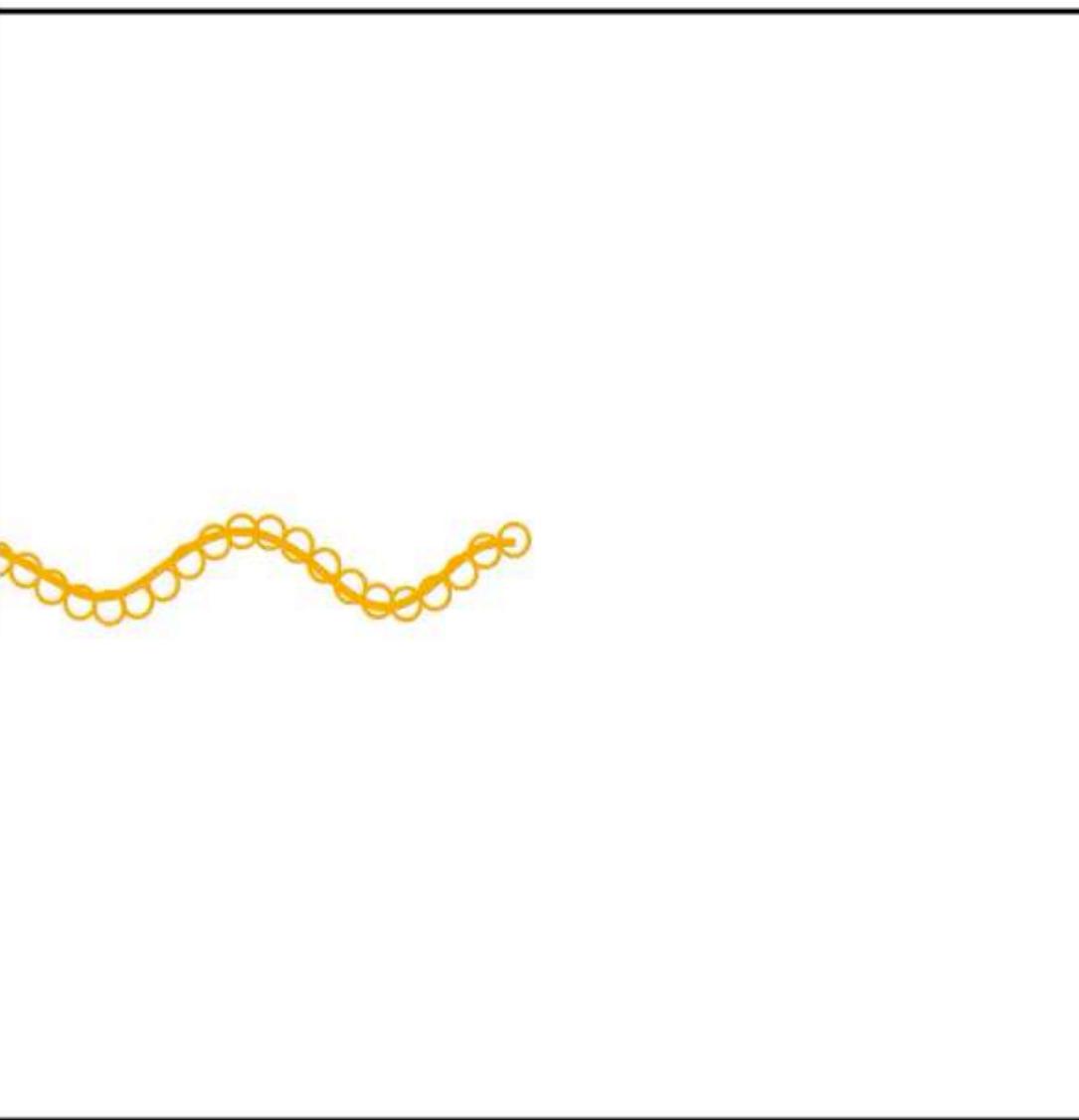
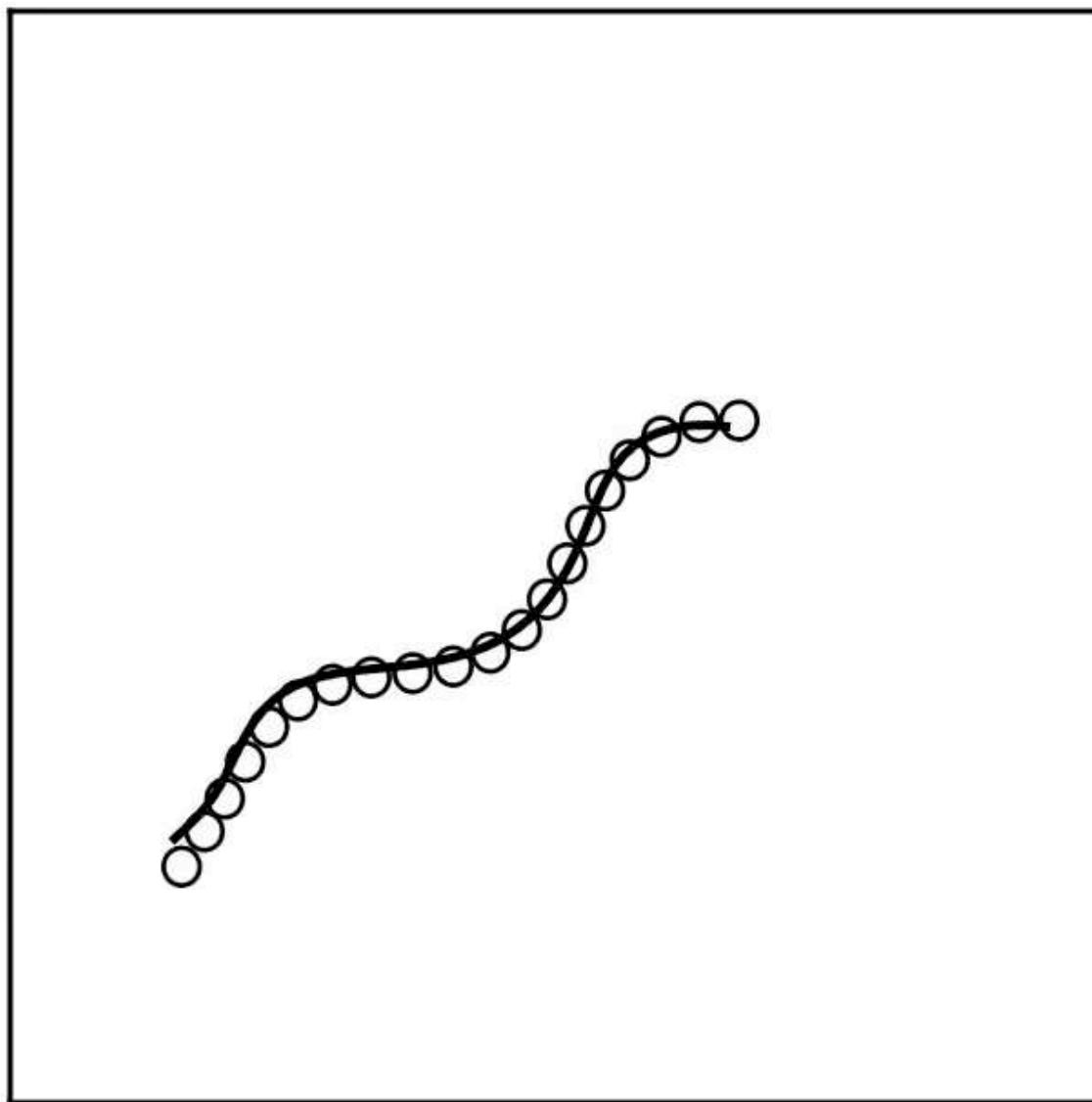
Flavell lab (MIT Picower)



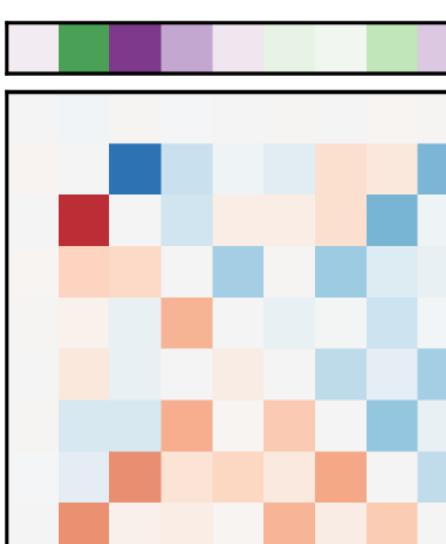
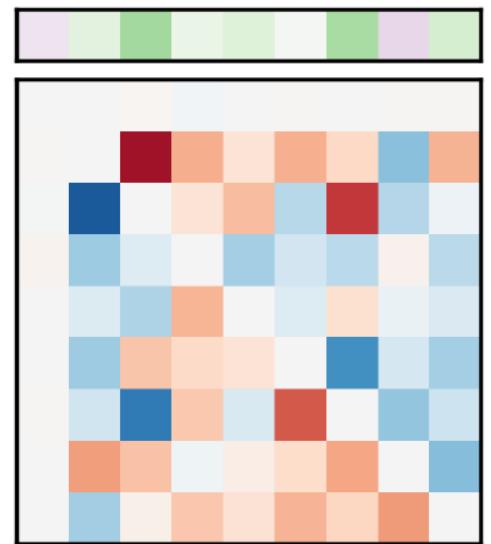
Goldman lab (Georgia Tech)



Alex's dorm room (MIT)



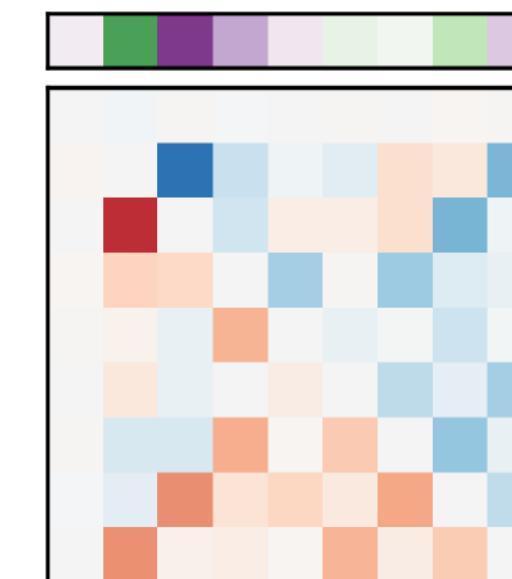
H



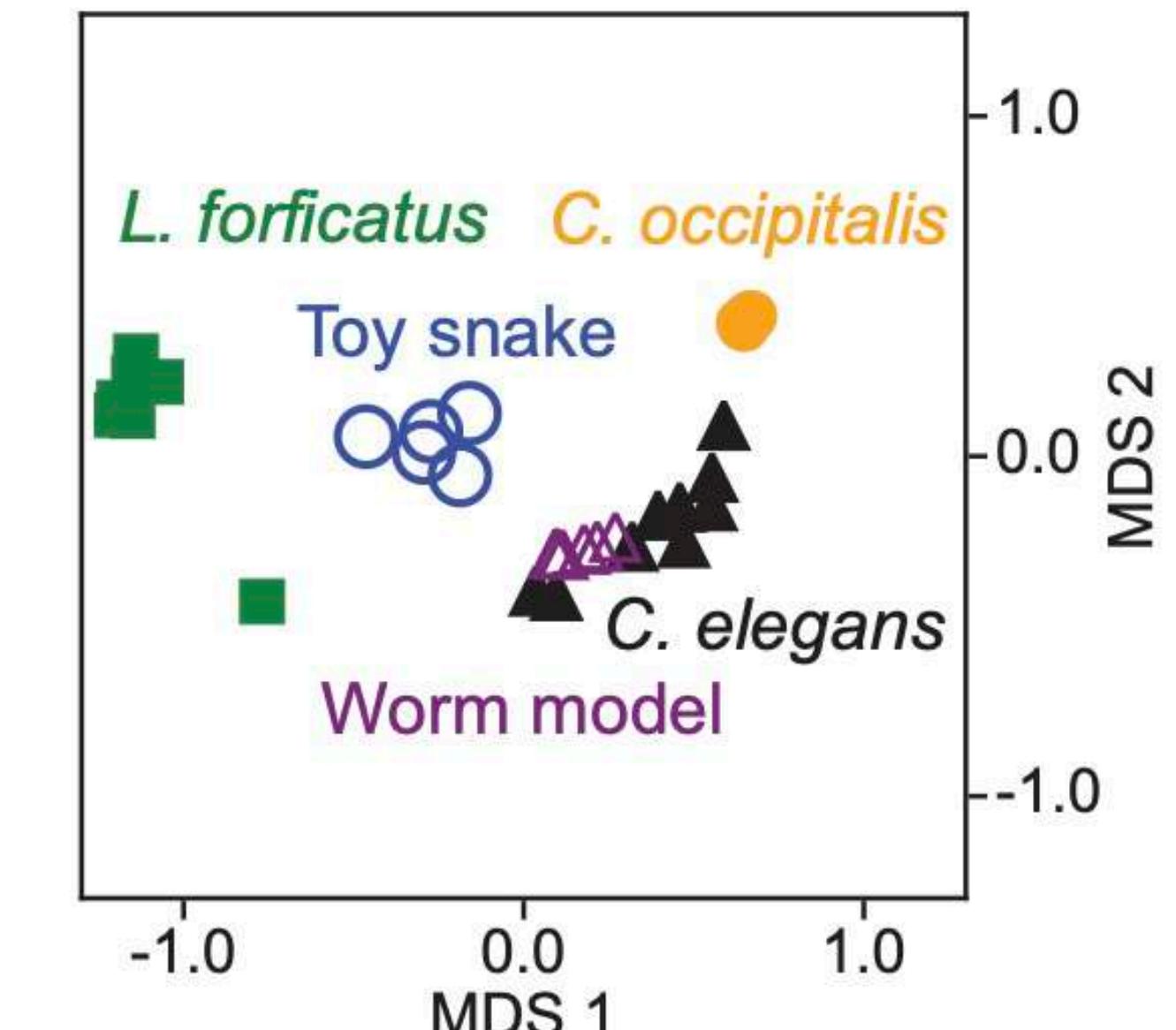
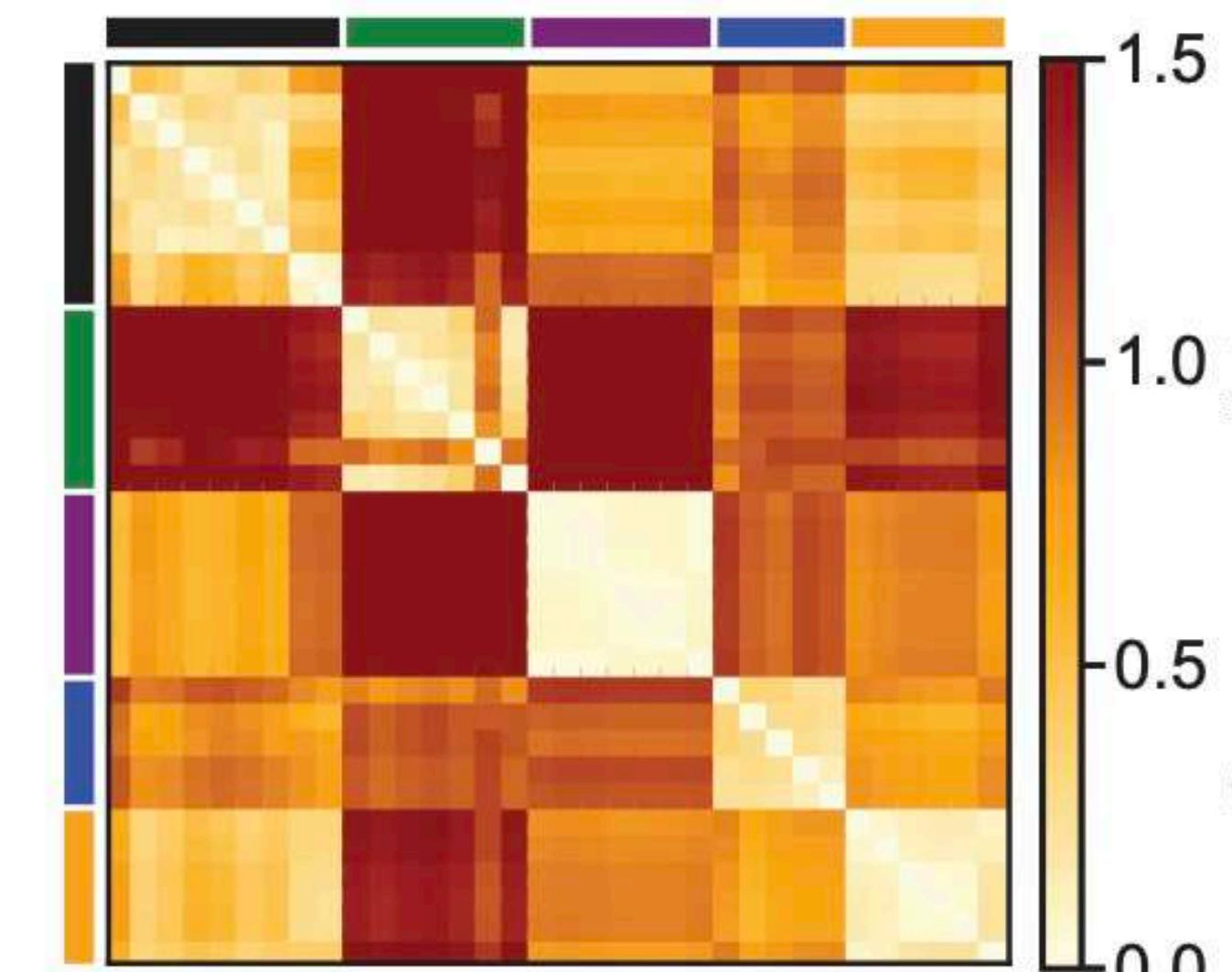
Schrödinger dynamics of undulatory locomotion

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

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

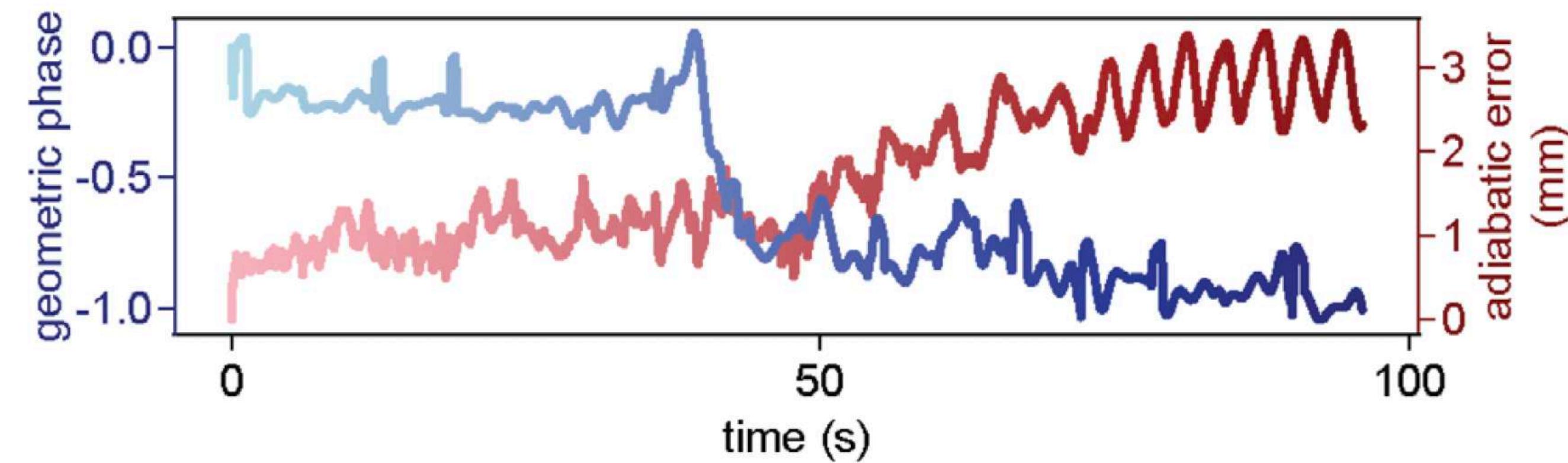
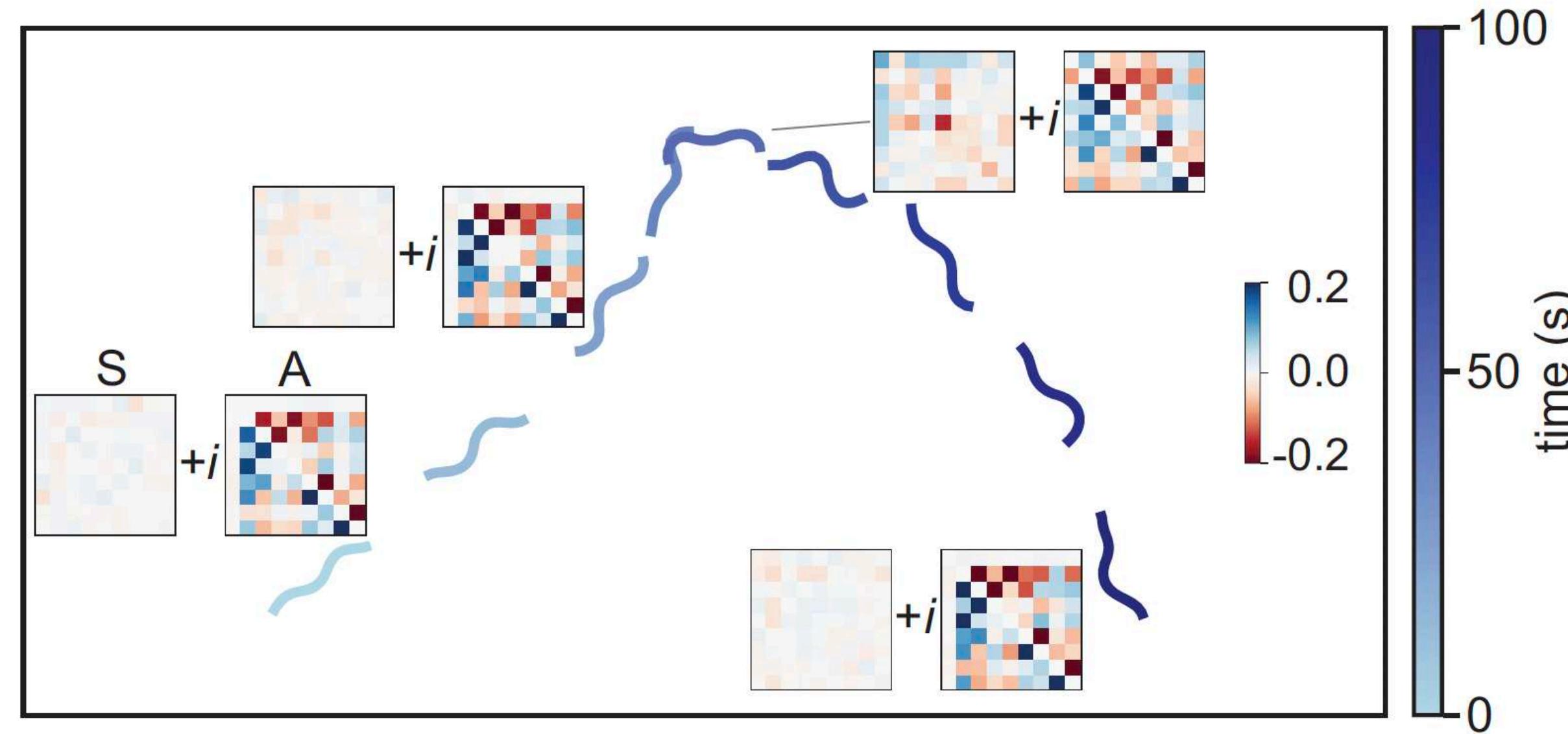


$$1 = \psi^\dagger \psi$$

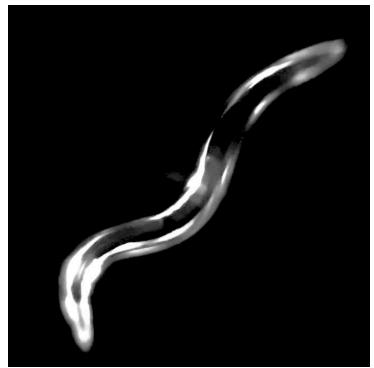


behavioral state \Leftrightarrow low-rank Hamiltonian \Rightarrow interspecies comparison

Time-dependent Hamiltonian and non-adiabatic turning



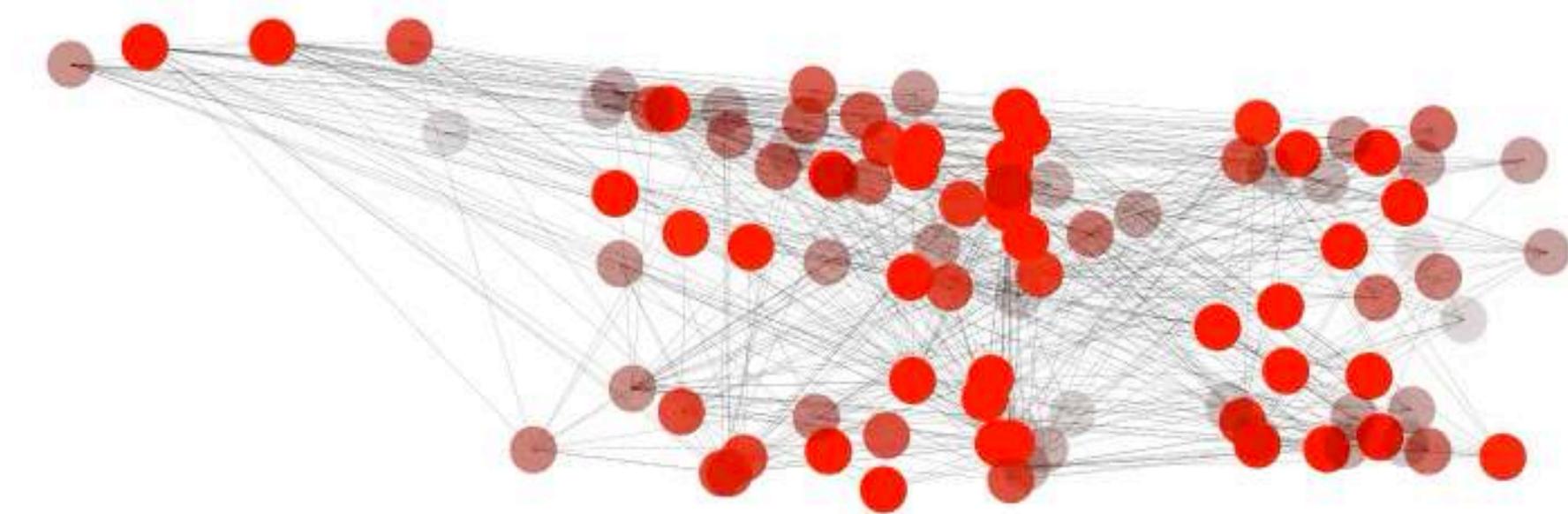
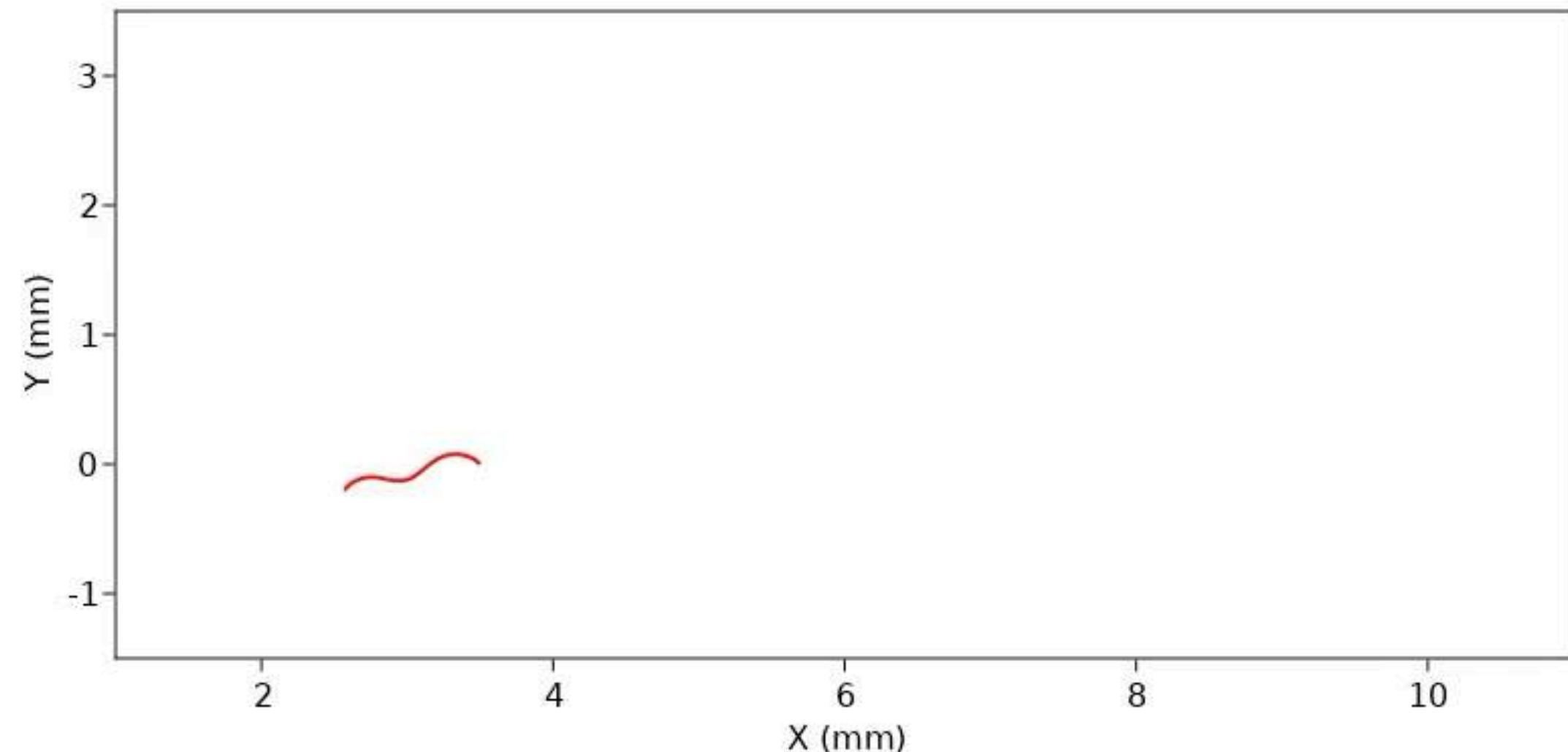
Outlook : Nonlinear model inference



C elegans

Spatiotemporal neuron activity \leftrightarrow physical dynamics

Phys Rev Lett 2023



Cohen et al, preprint 2025 (available on request)

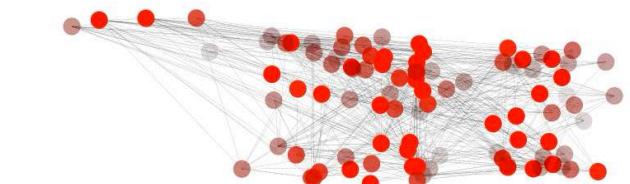
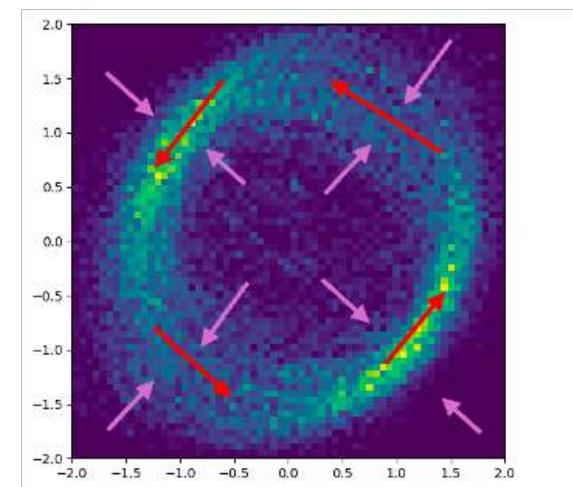


Steven Flavell



Neuromechanics

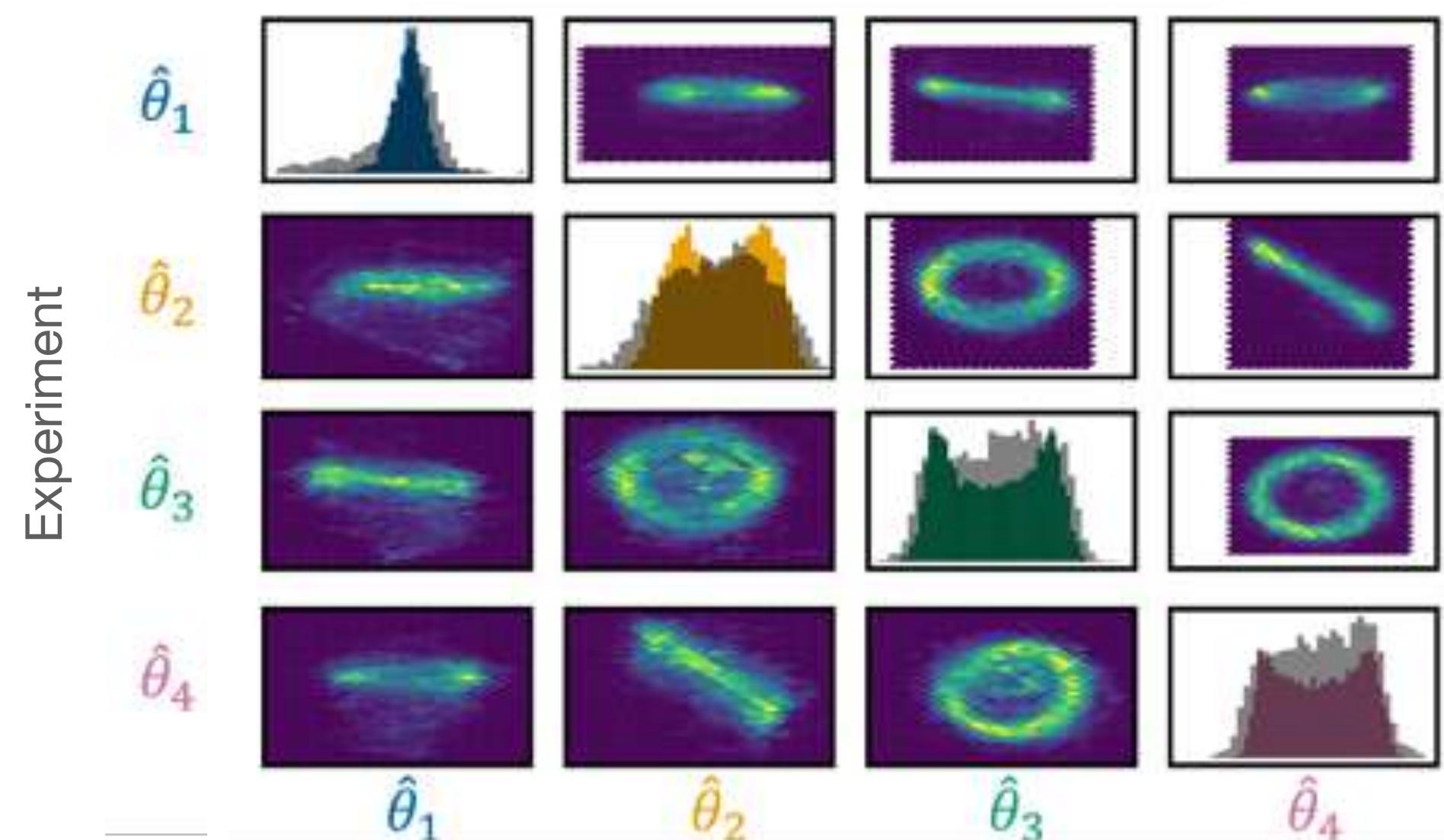
Outlook : Nonlinear SDE inference



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

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

Nonlinear SDE model

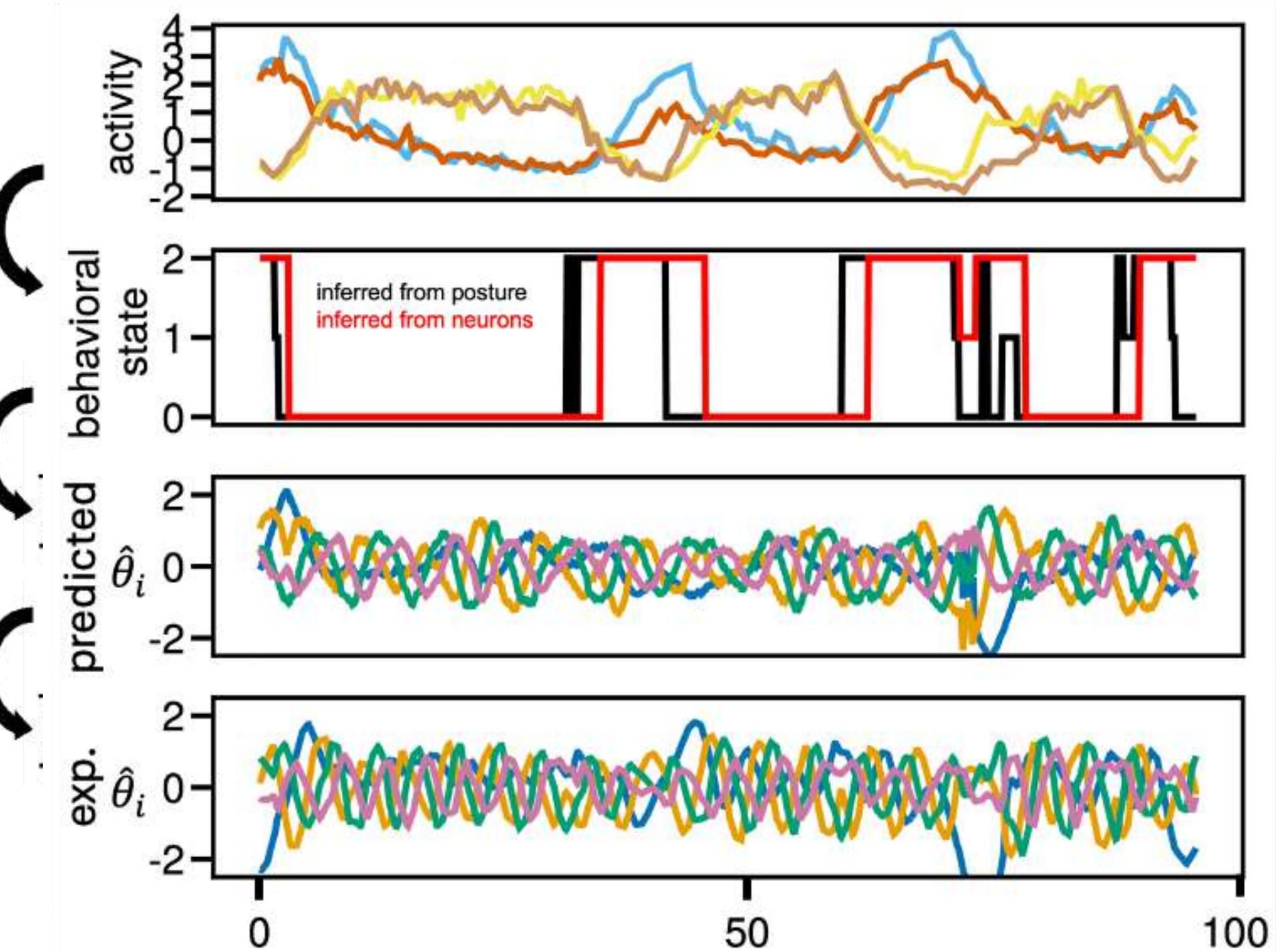


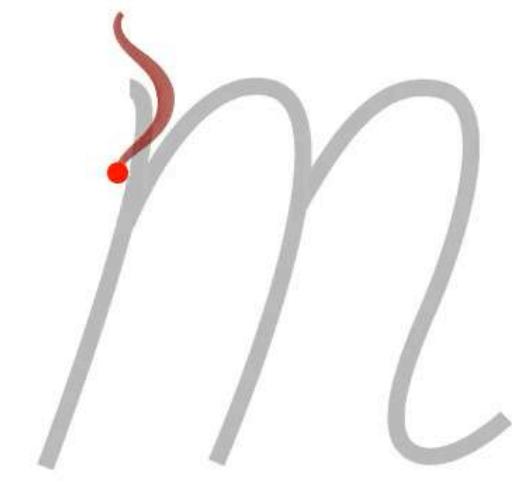
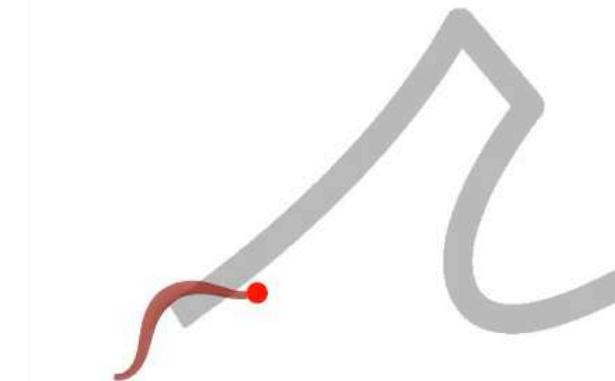
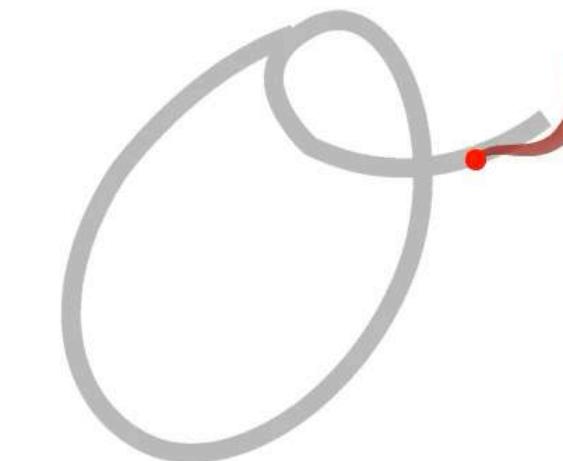
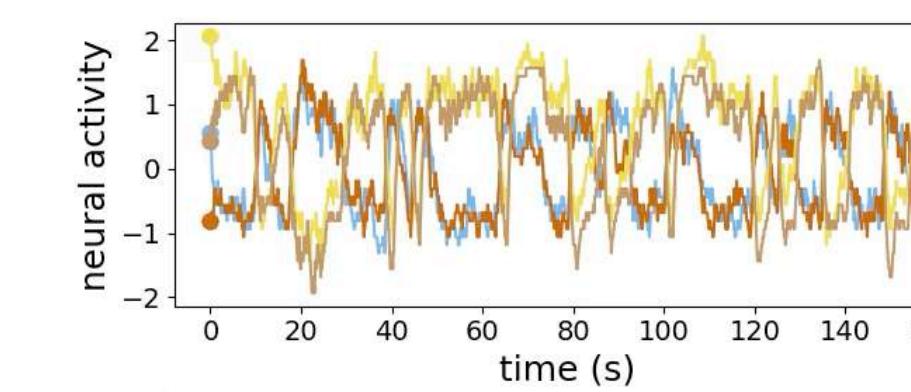
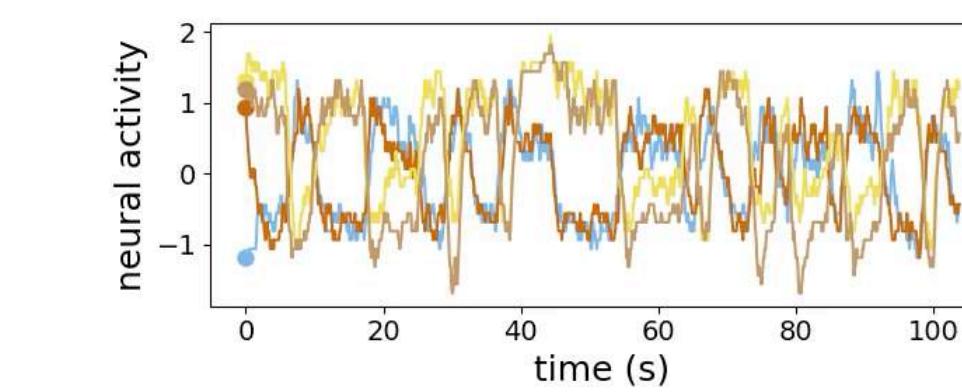
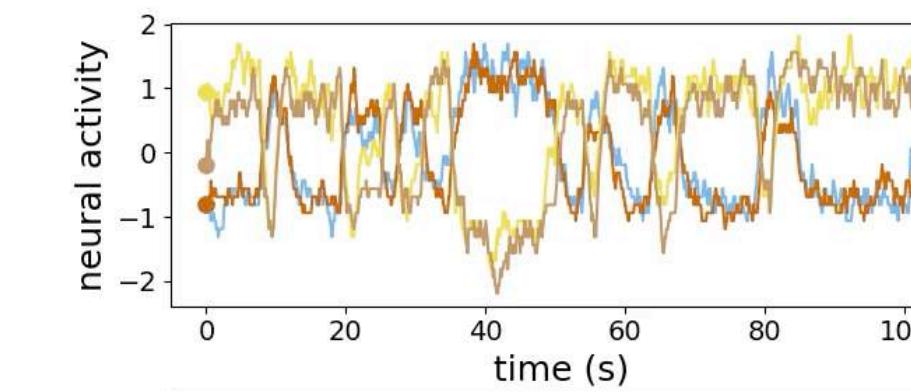
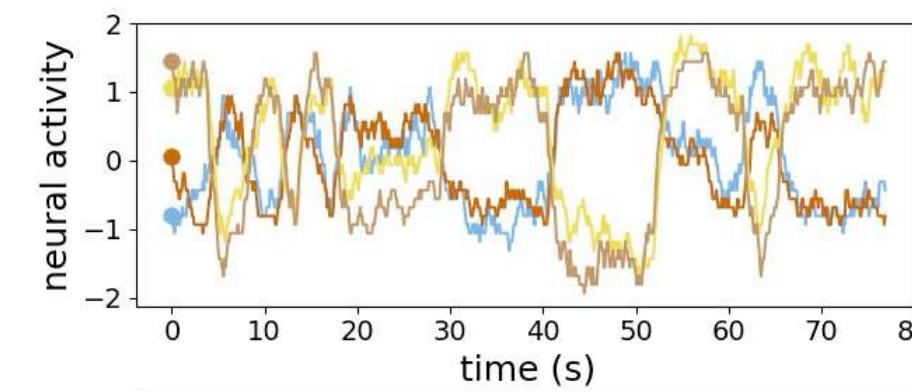
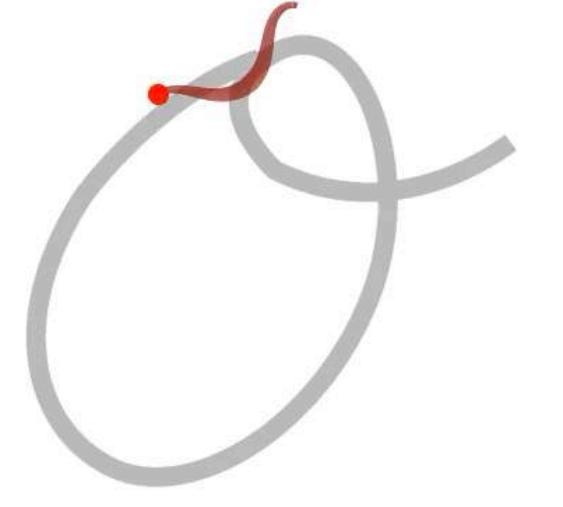
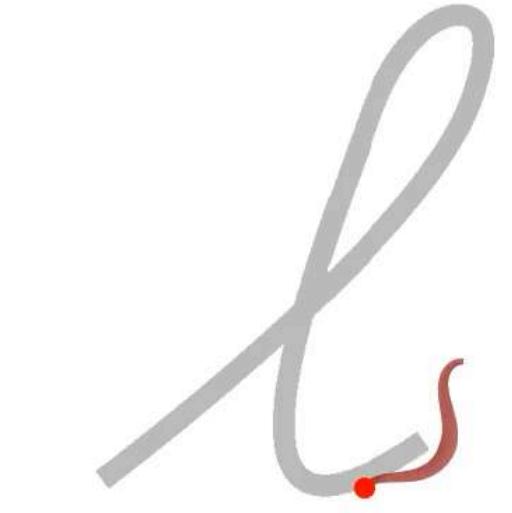
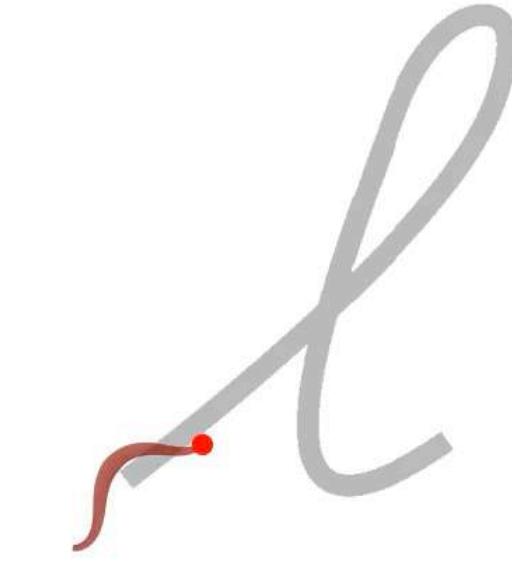
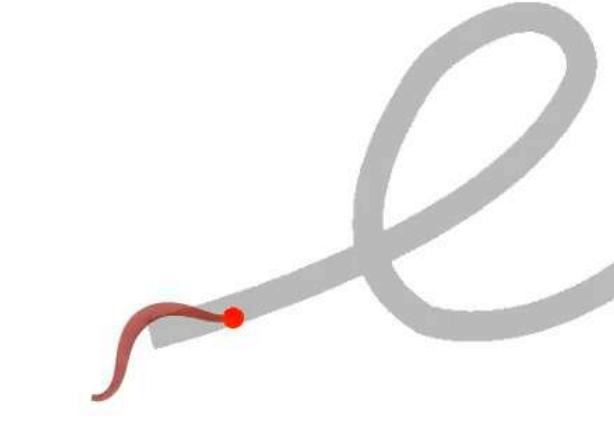
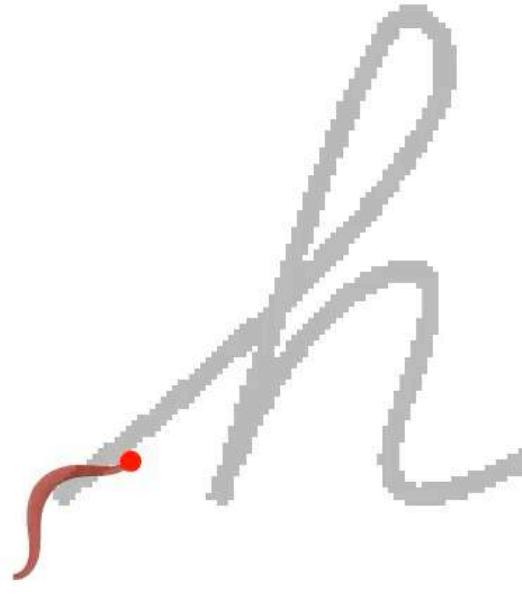
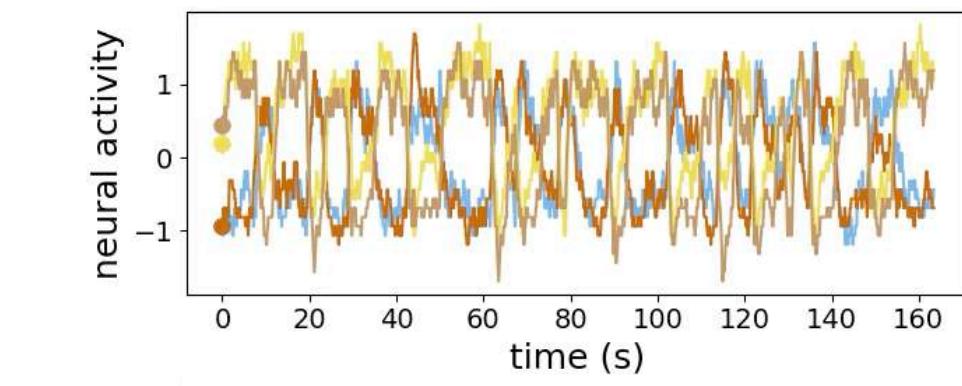
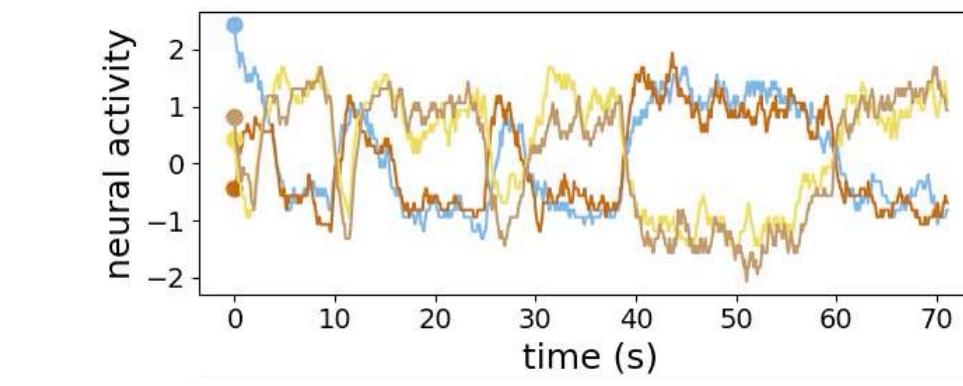
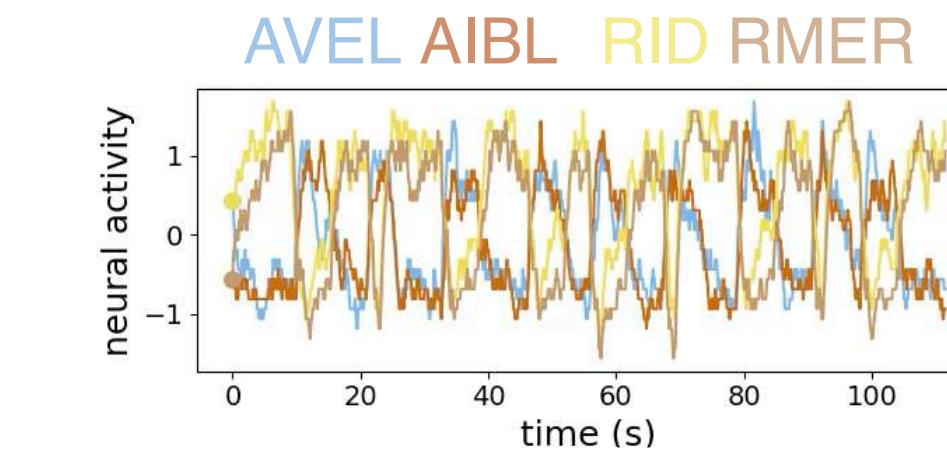
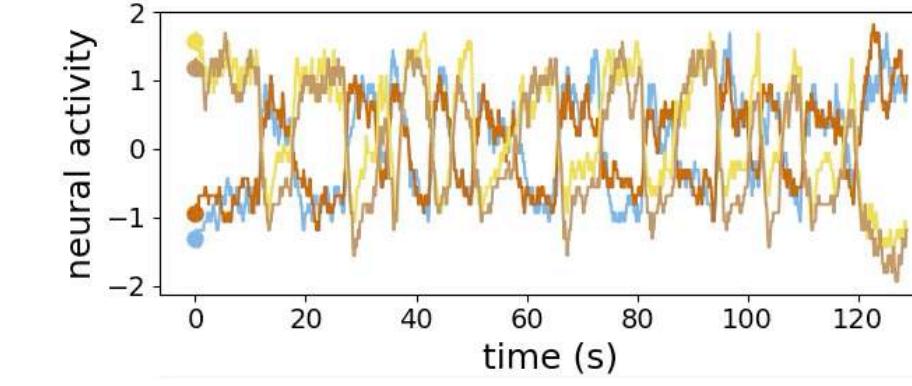
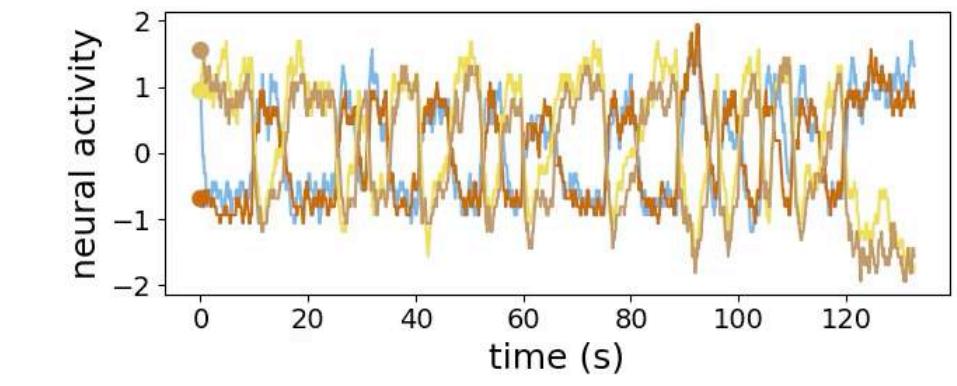
infer behavior
from neural
activity

prediction
motion from
behavior

compare with
measured
motion

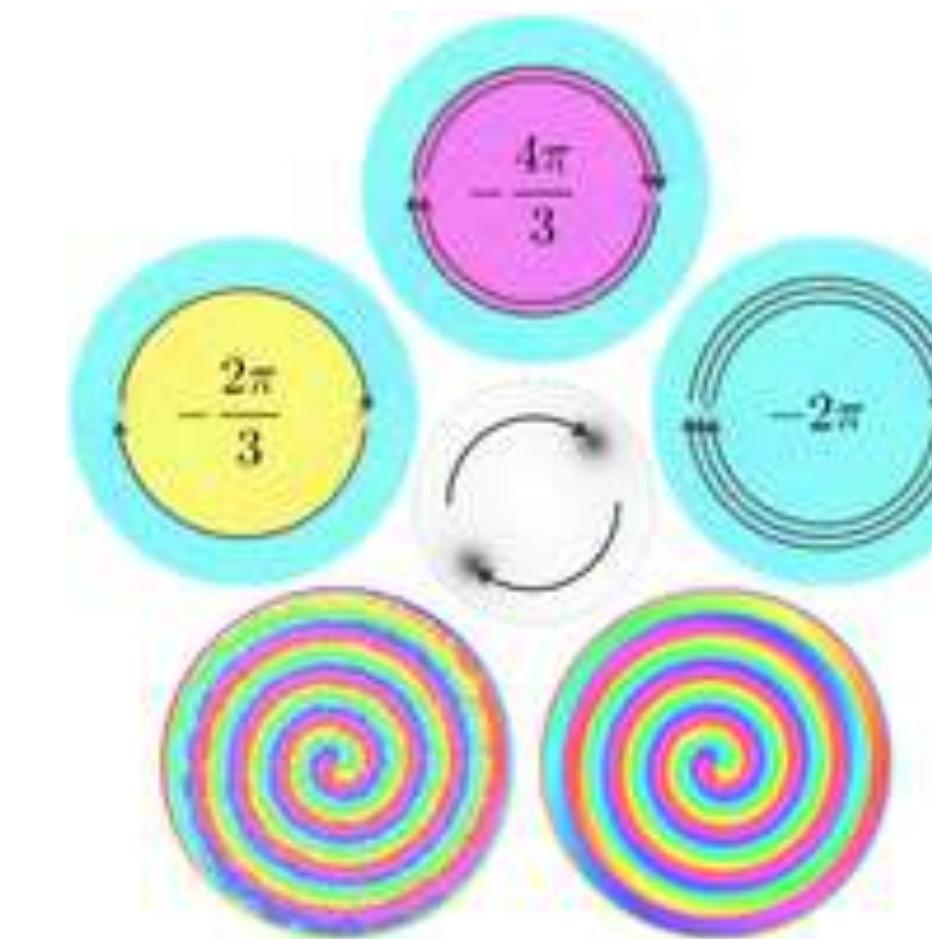
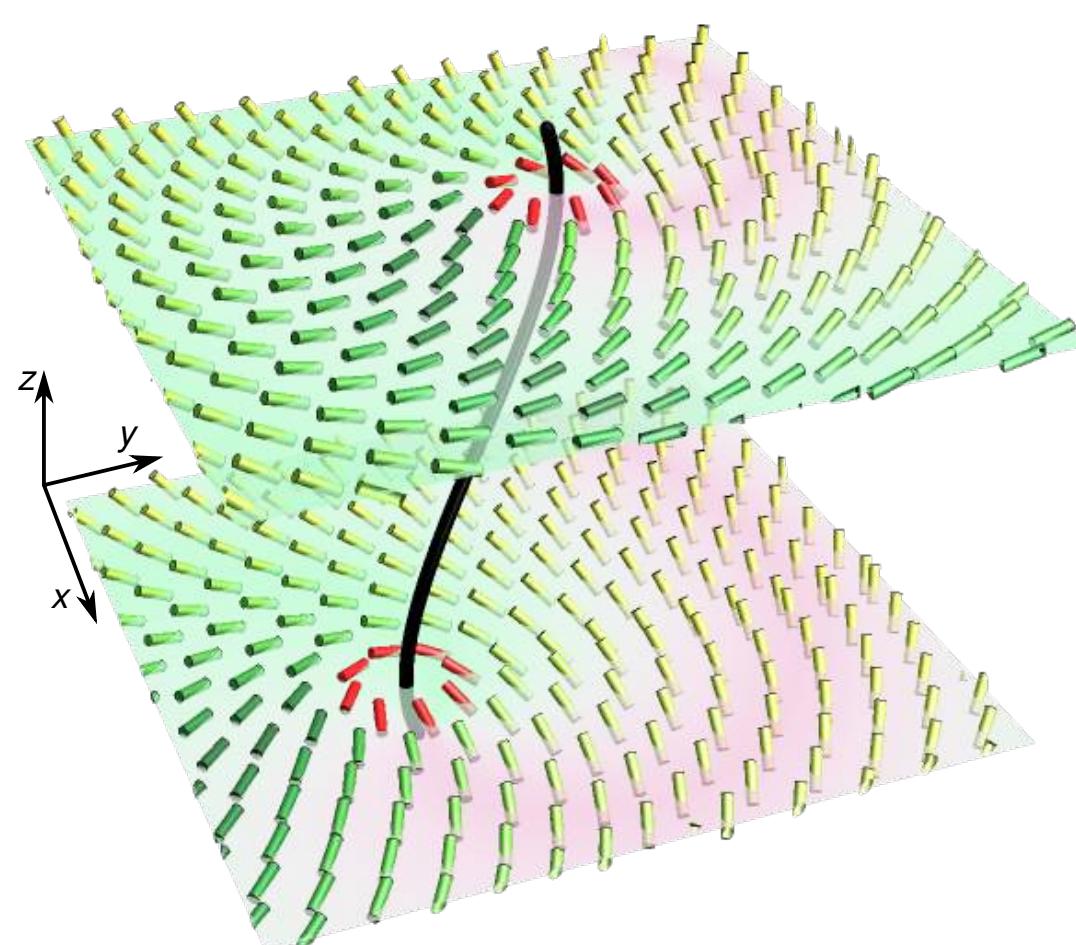
Predicting shape-dynamics from neuron activity





Part 2:

Topological defects & computation in liquid crystals

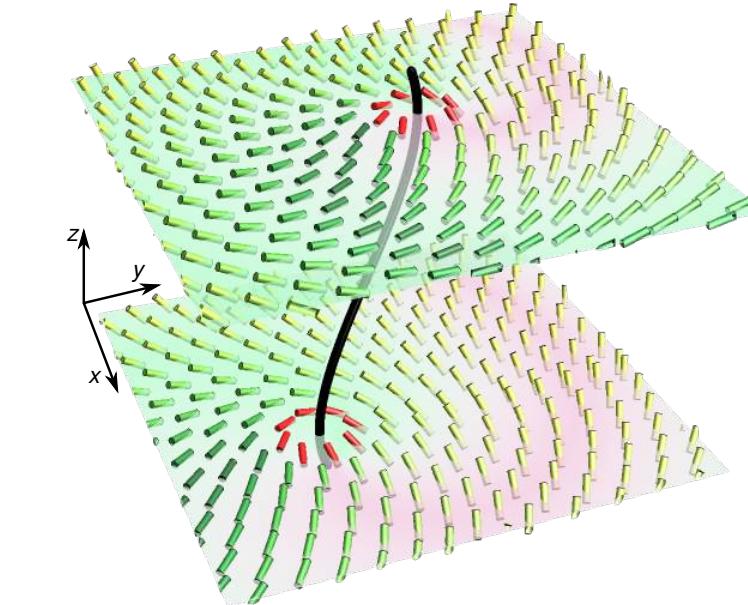


Science Advances 2022

PRX 2022

Broader motivation

Can we use topological defects as computational ‘bits’ ?



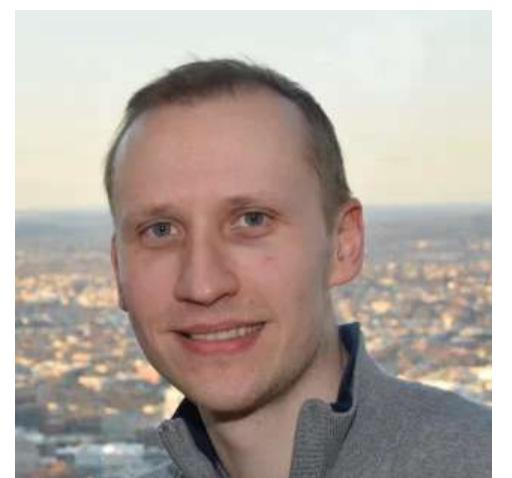
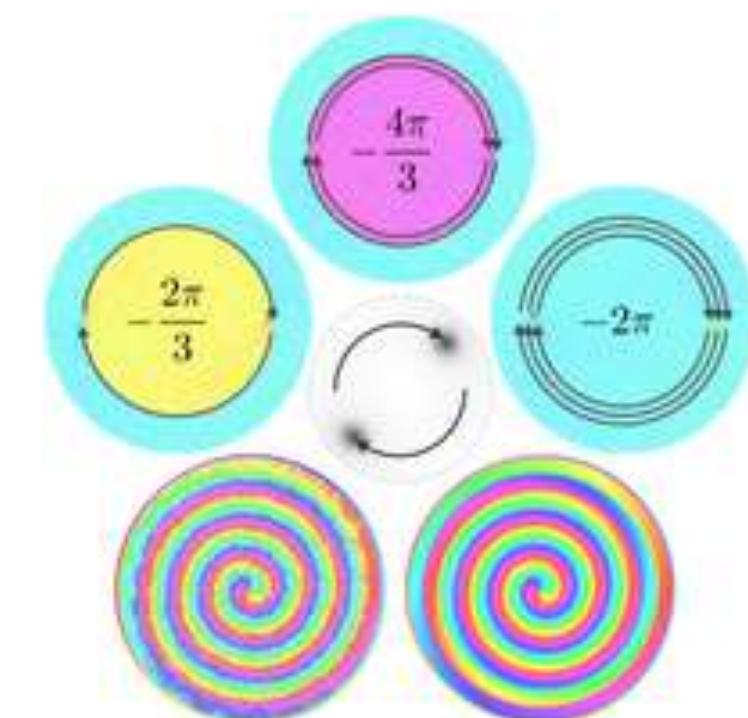
Ziga Kos
(MIT \Rightarrow Ljubljana)

Science Advances 2022

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 ?

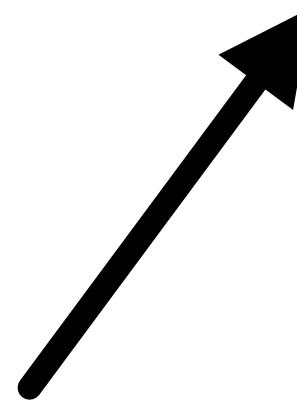


Alex Mietke
(MIT \Rightarrow Oxford)

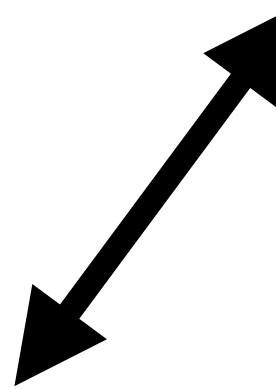
PRX 2022

Dihedral particle symmetry

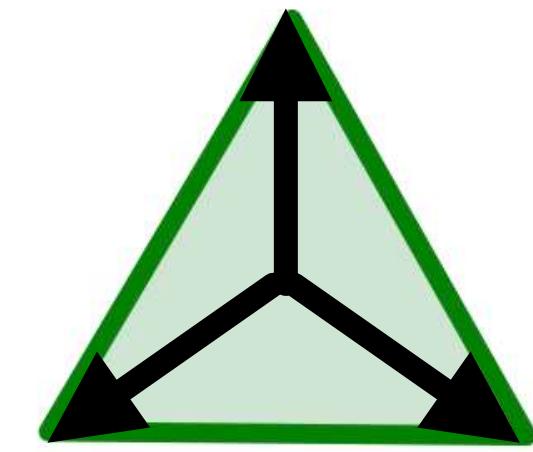
$k = 1$



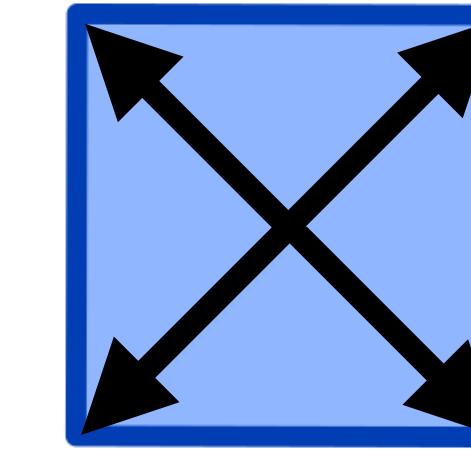
$k = 2$



$k = 3$



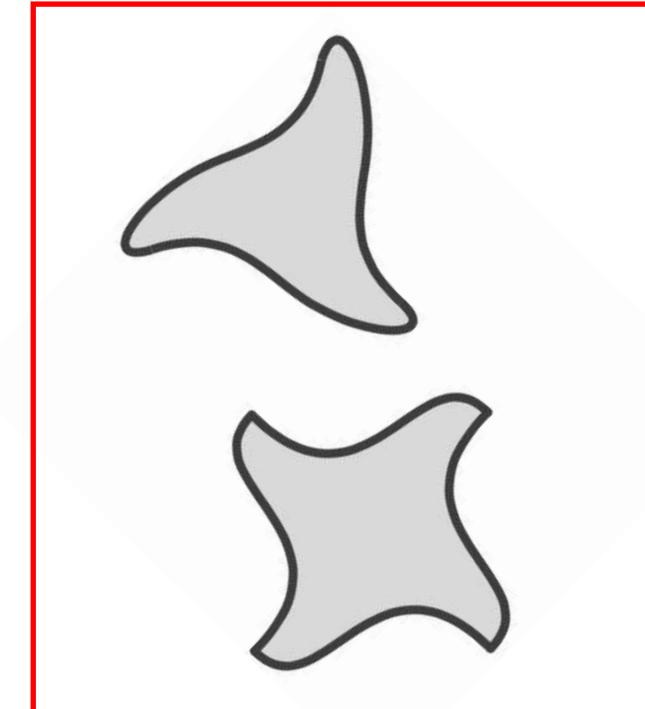
$k = 4$



etc.

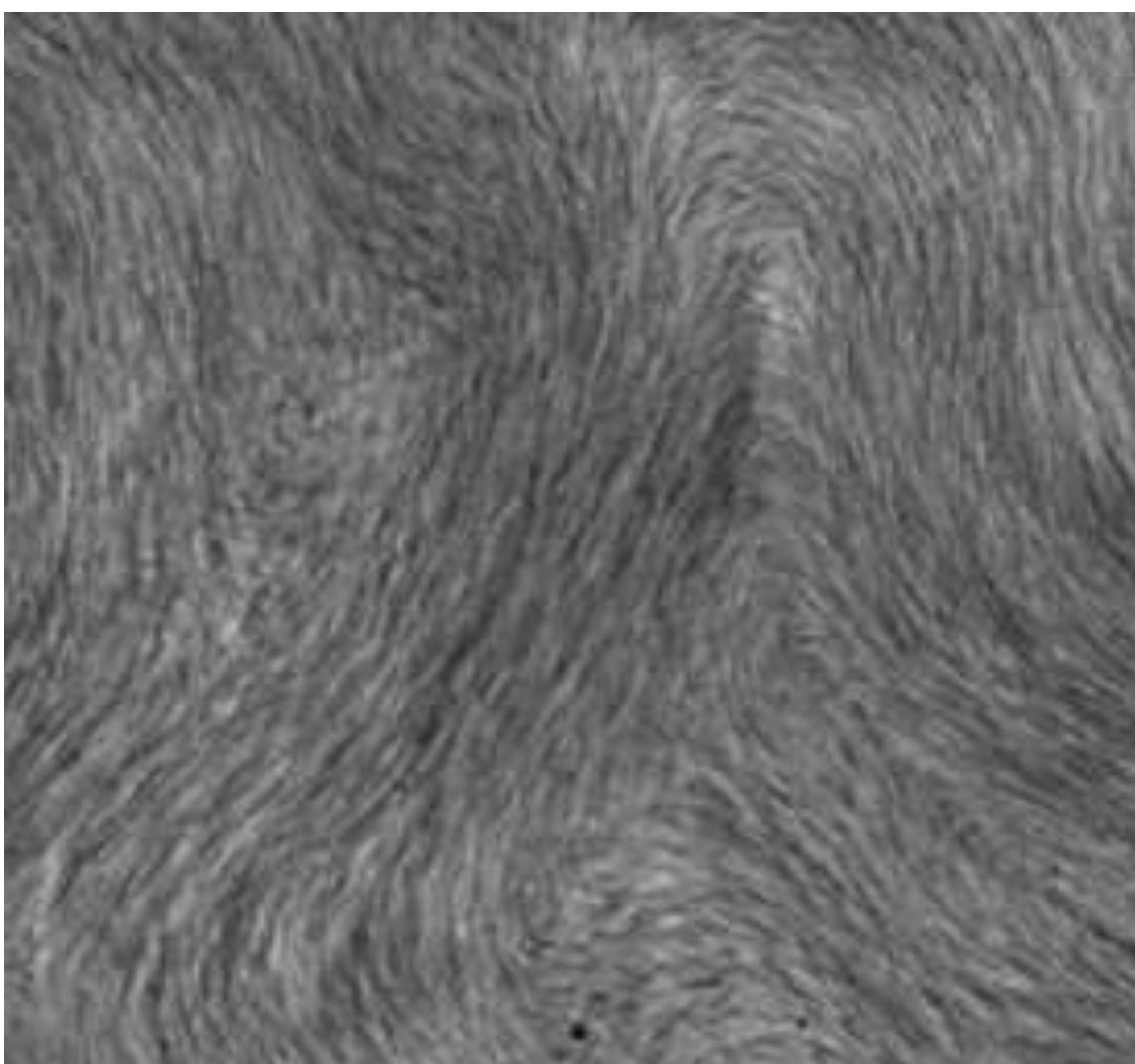
Symmetric under $\frac{2\pi}{k}$ rotation
~~+ reflection symmetries~~

Dihedral group $C_n D_n$



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

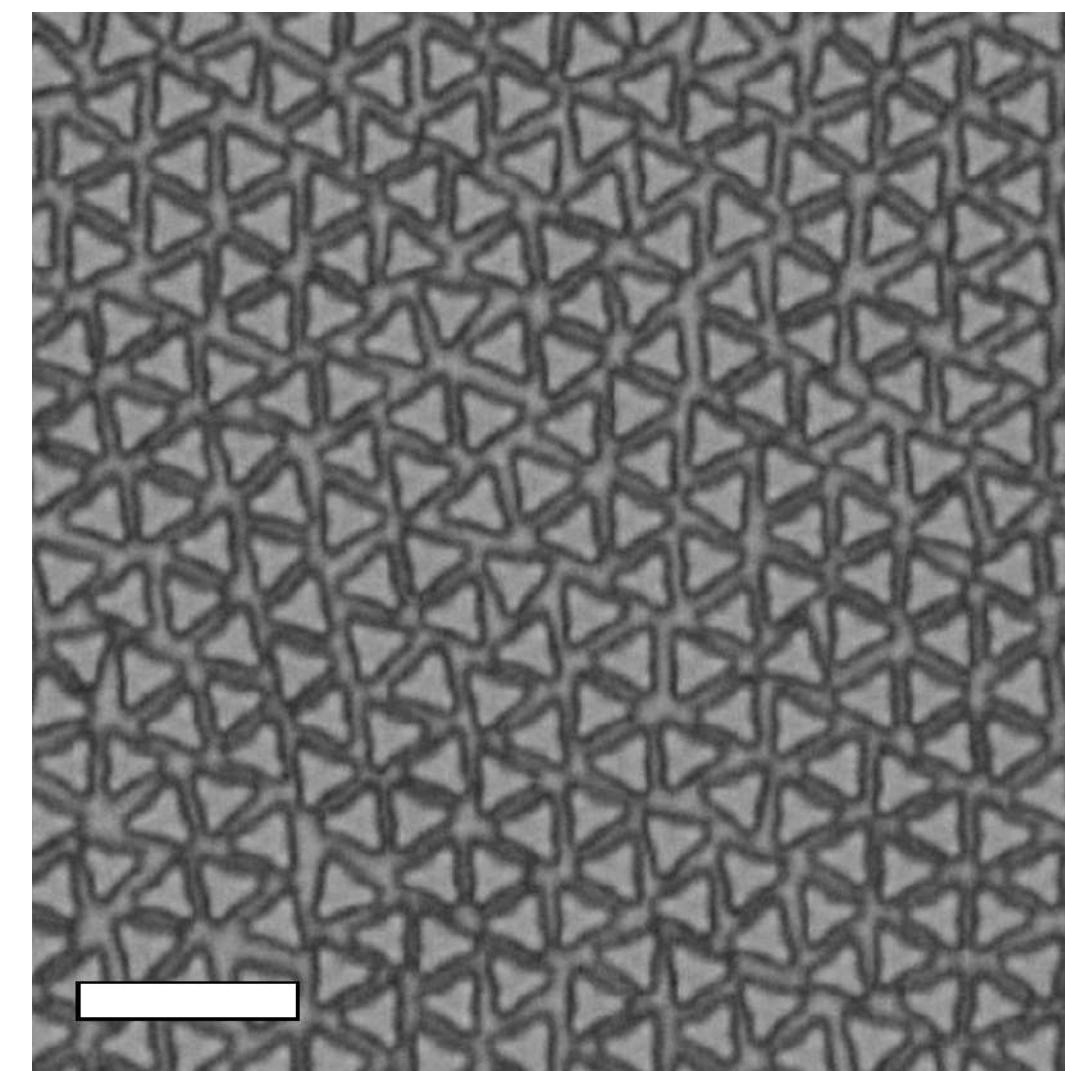
2-atic
(nematic)



Zhou et al. PNAS 2014

bacteria

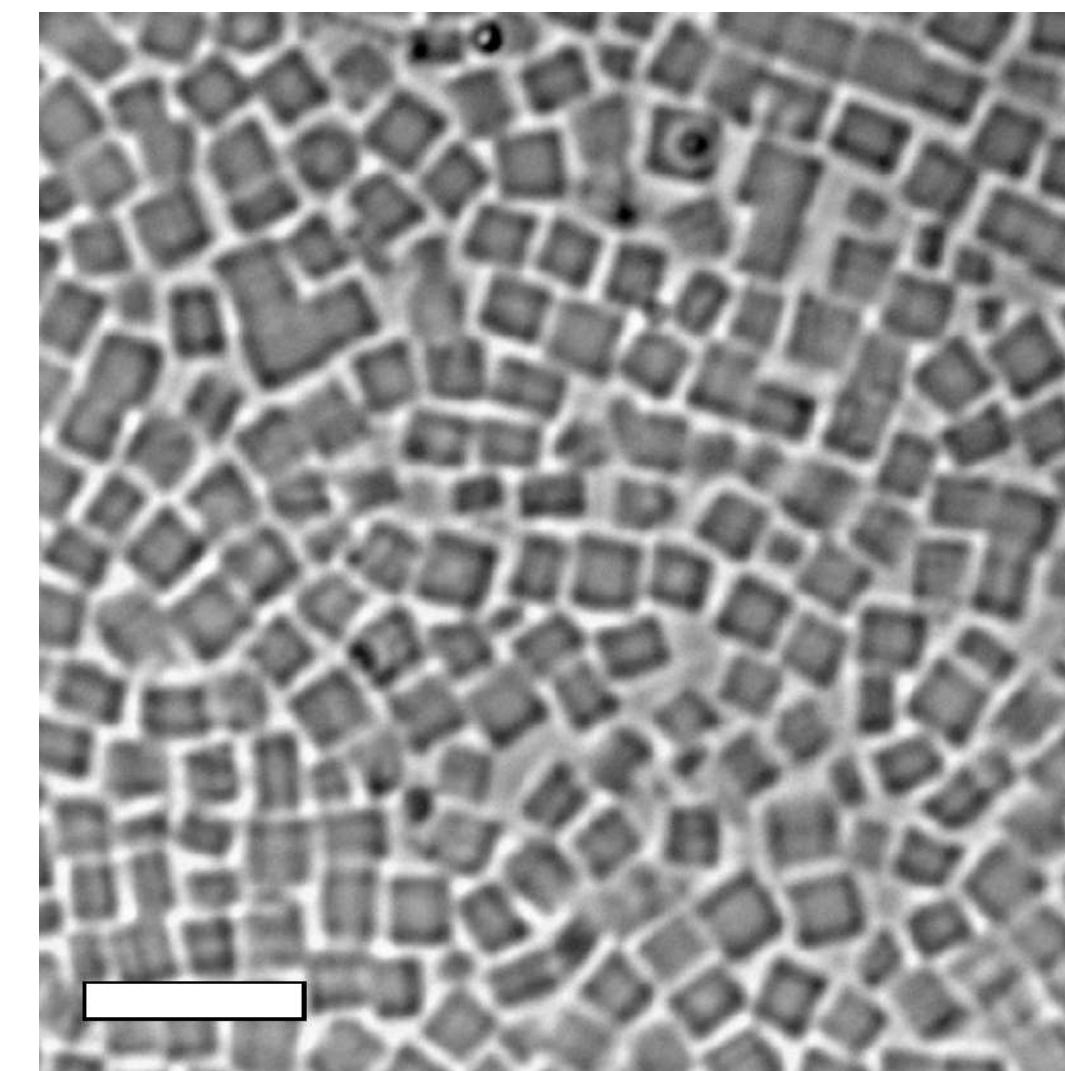
3-atic



Zhao et al. Nat Comm 2012

colloids

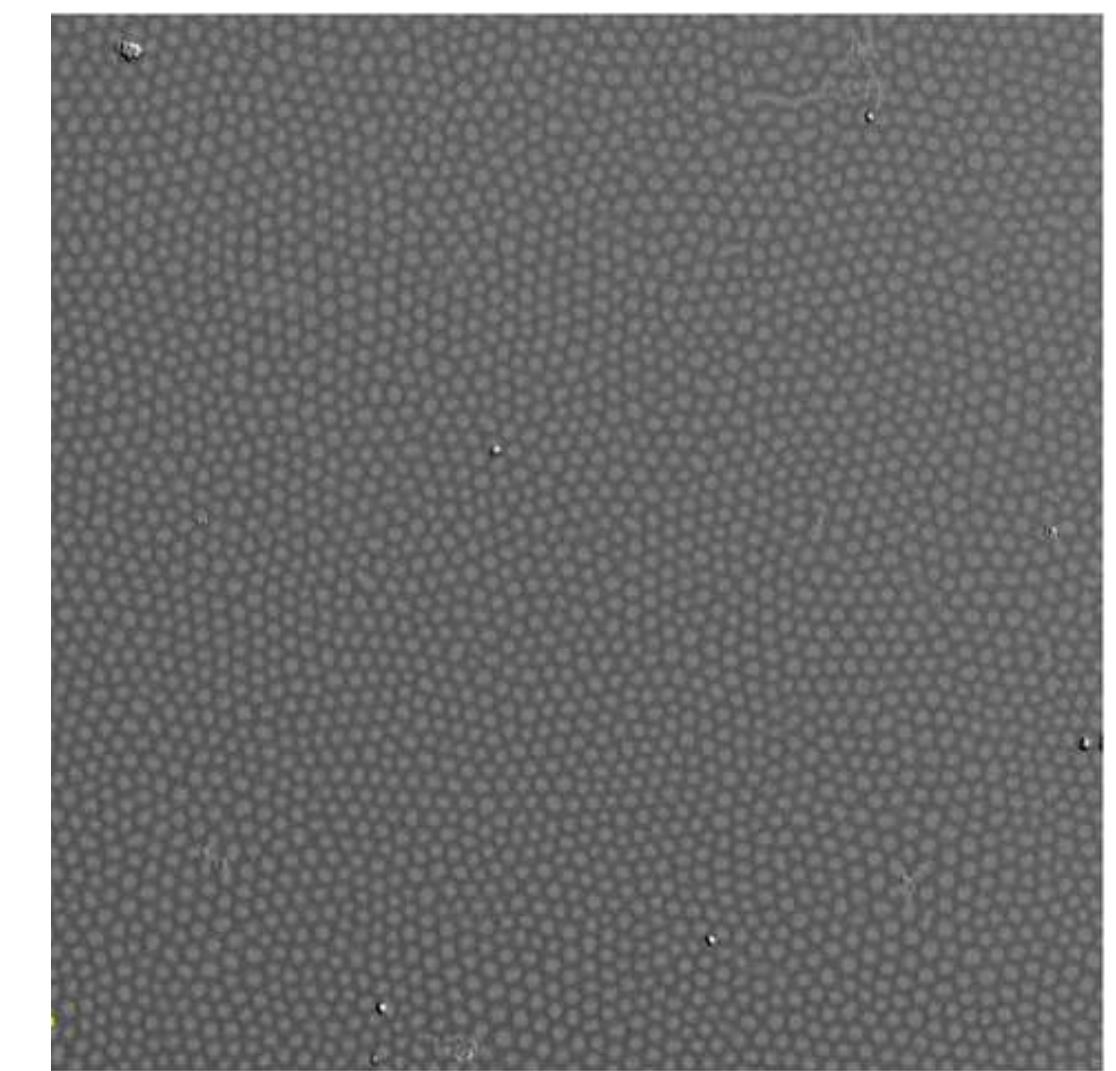
4-atic



Löffler. Thesis 2018 (Konstanz)

colloids

6-atic



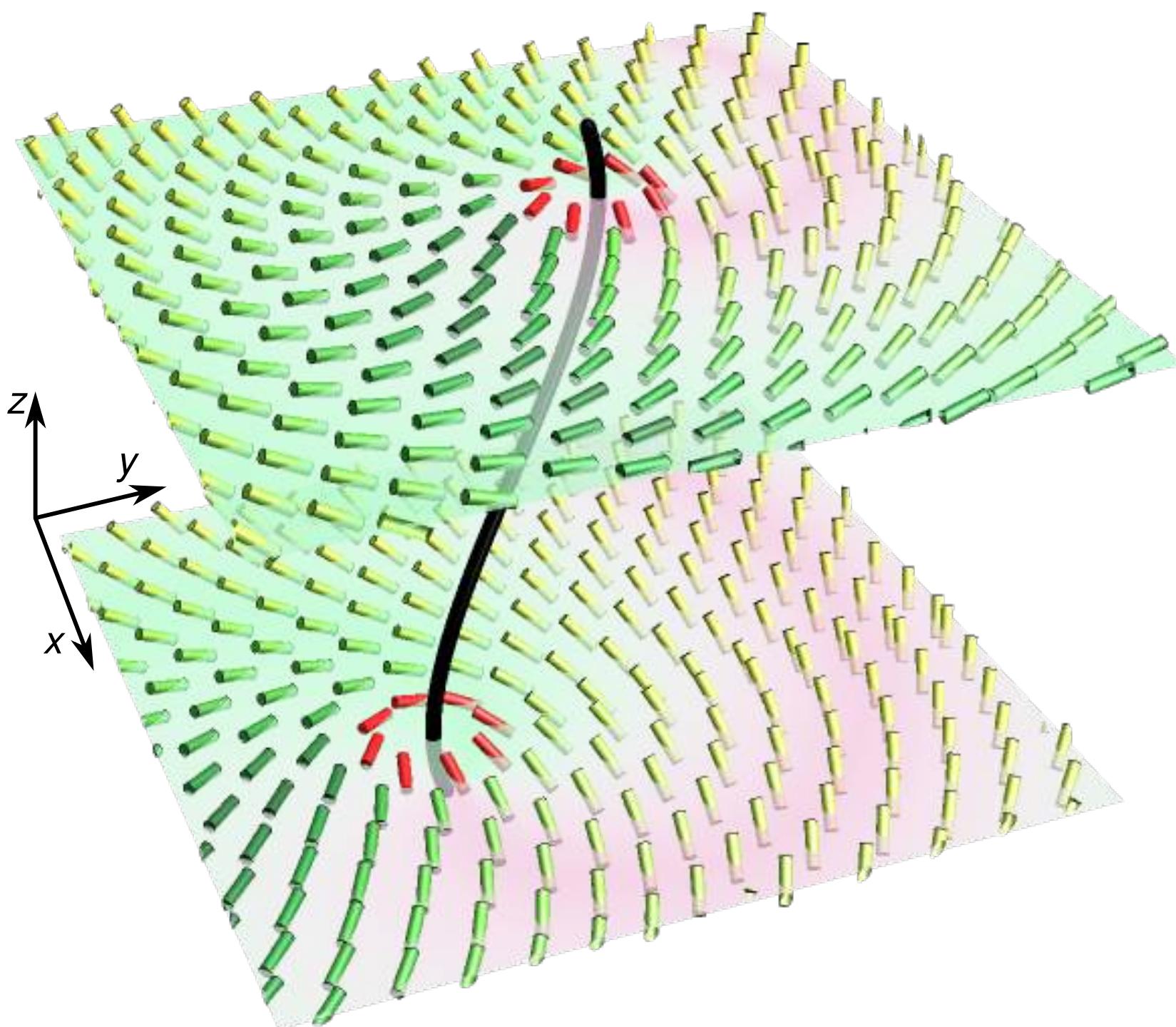
Zazvorka et al. Adv Funct Mater 2020

skyrmions

Positionally disordered but orientationally ordered

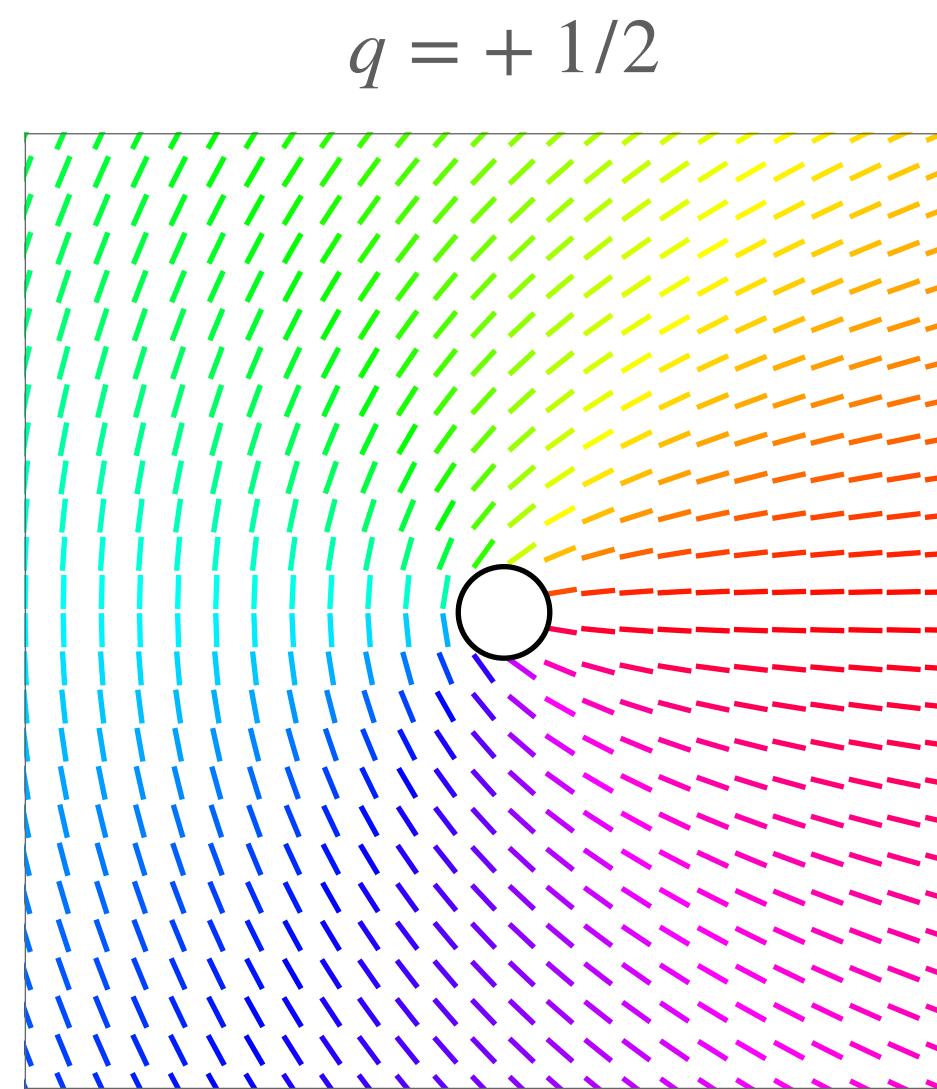
See also: Bowick & Giomi. Advances in Physics, 2009
Giomi, Toner & Sarkar. Phys Rev E 2022

Topological defects in nematic liquid crystals

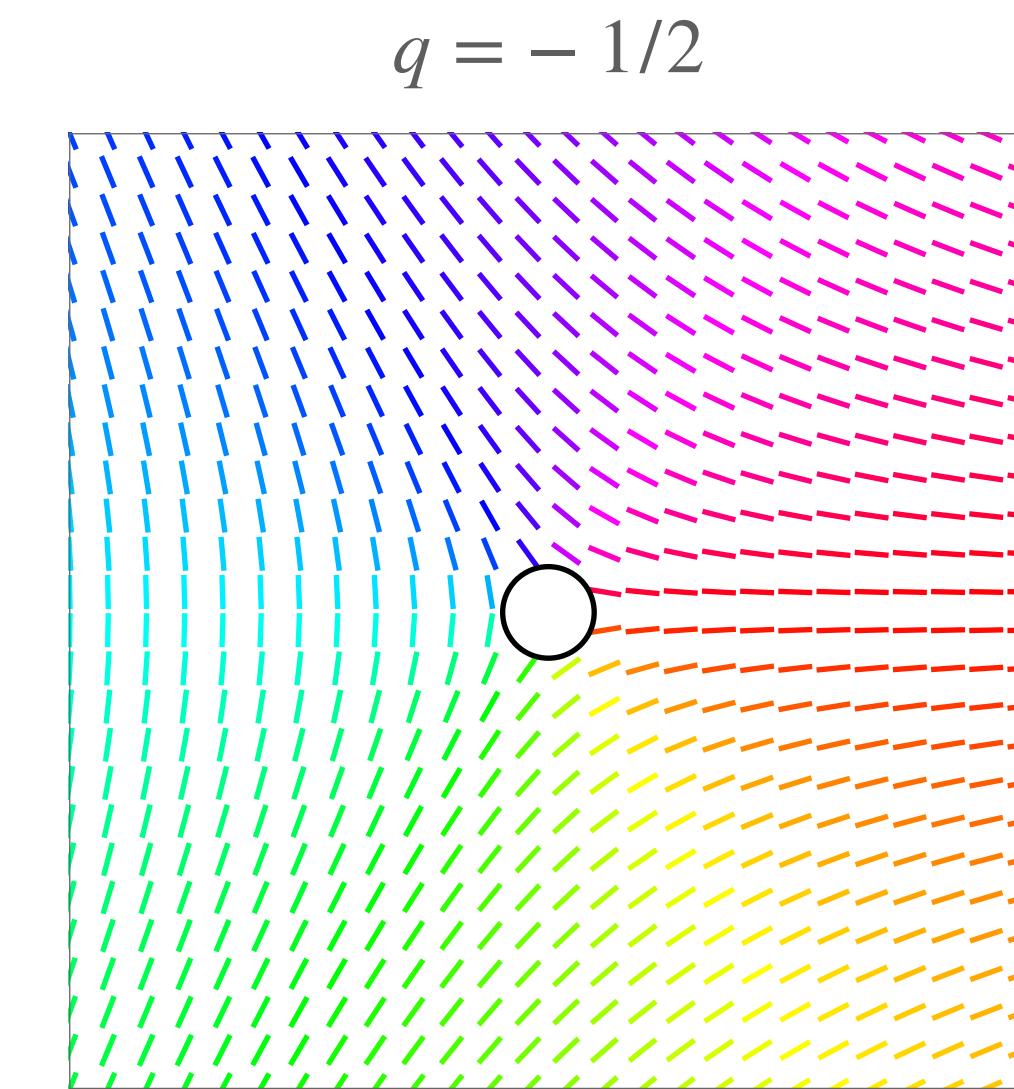


Uni-axial nematic LC

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



$q = +1/2$

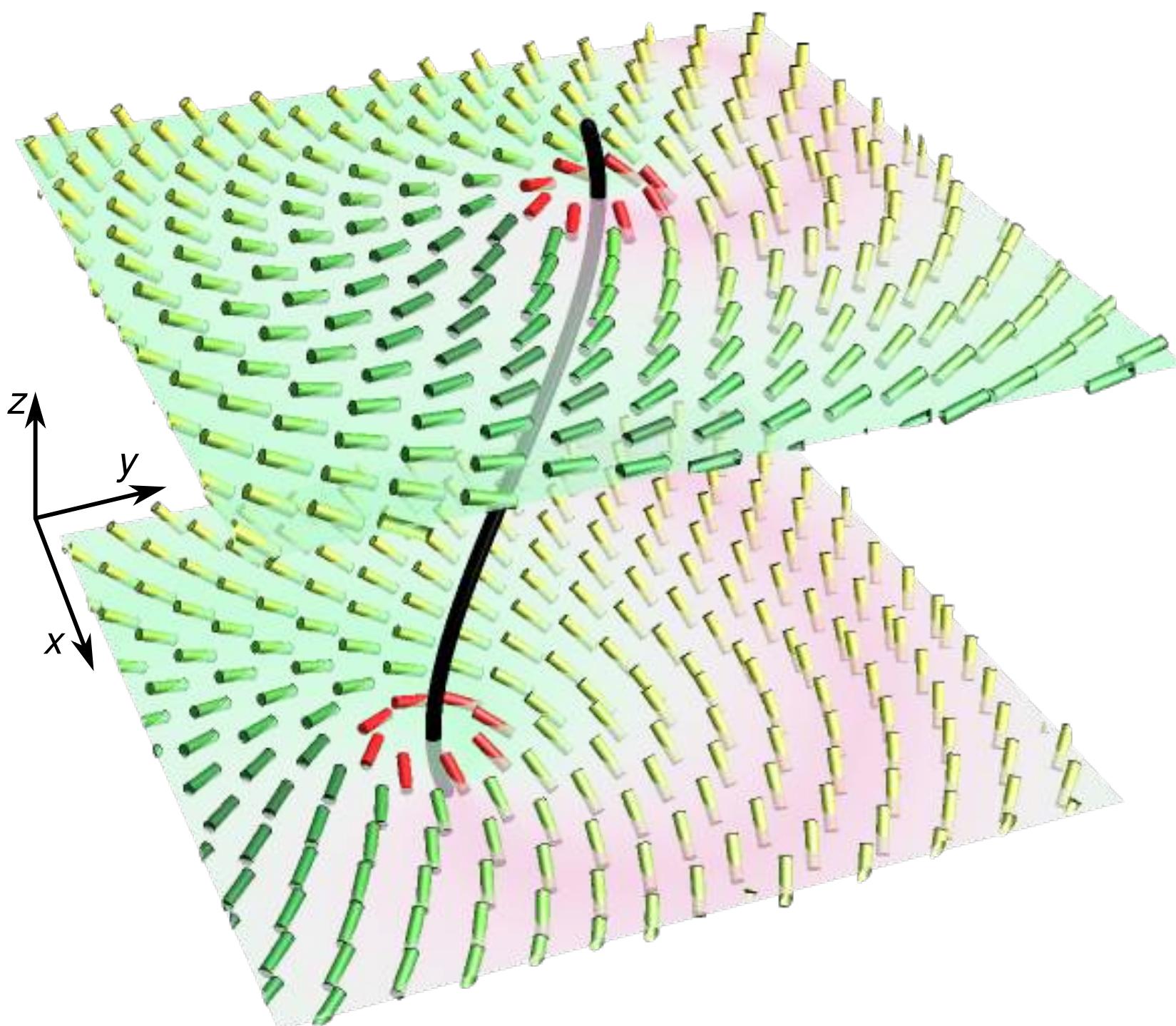


$q = -1/2$

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

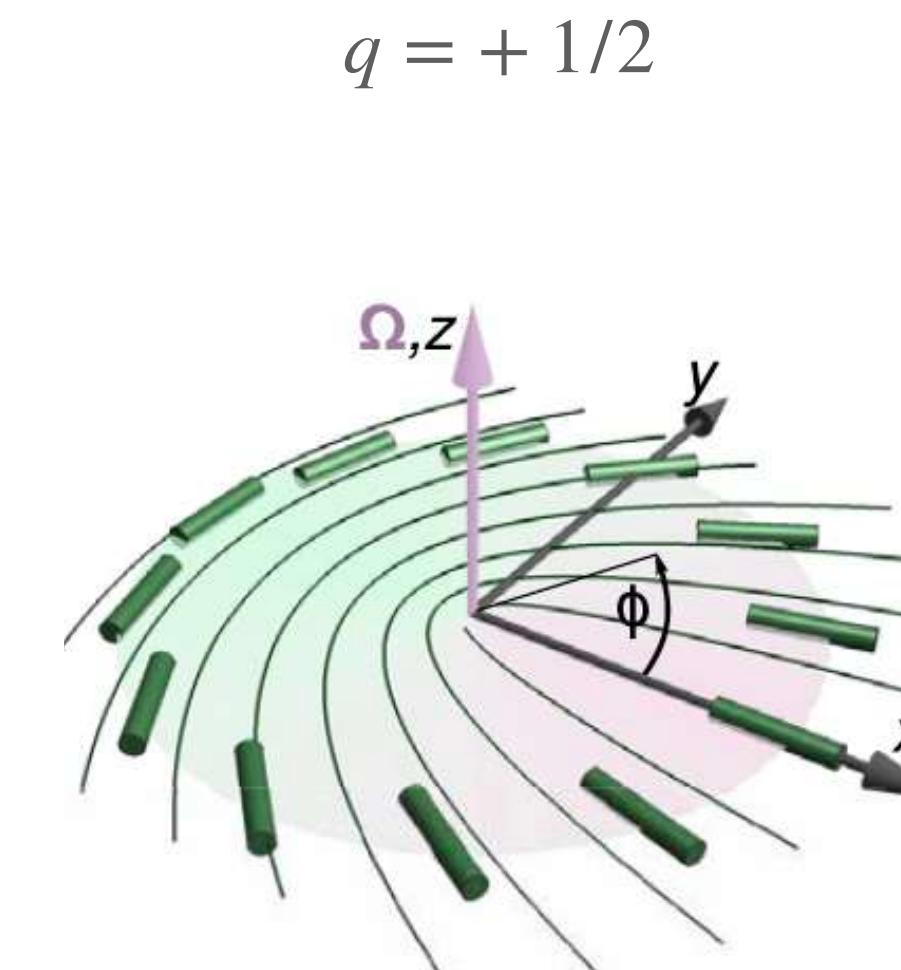
$$\mathbf{n}_-(\mathbf{r}) = (\cos \frac{\phi}{2}, \sin \frac{-\phi}{2}, 0)$$

Topological defects in nematic liquid crystals

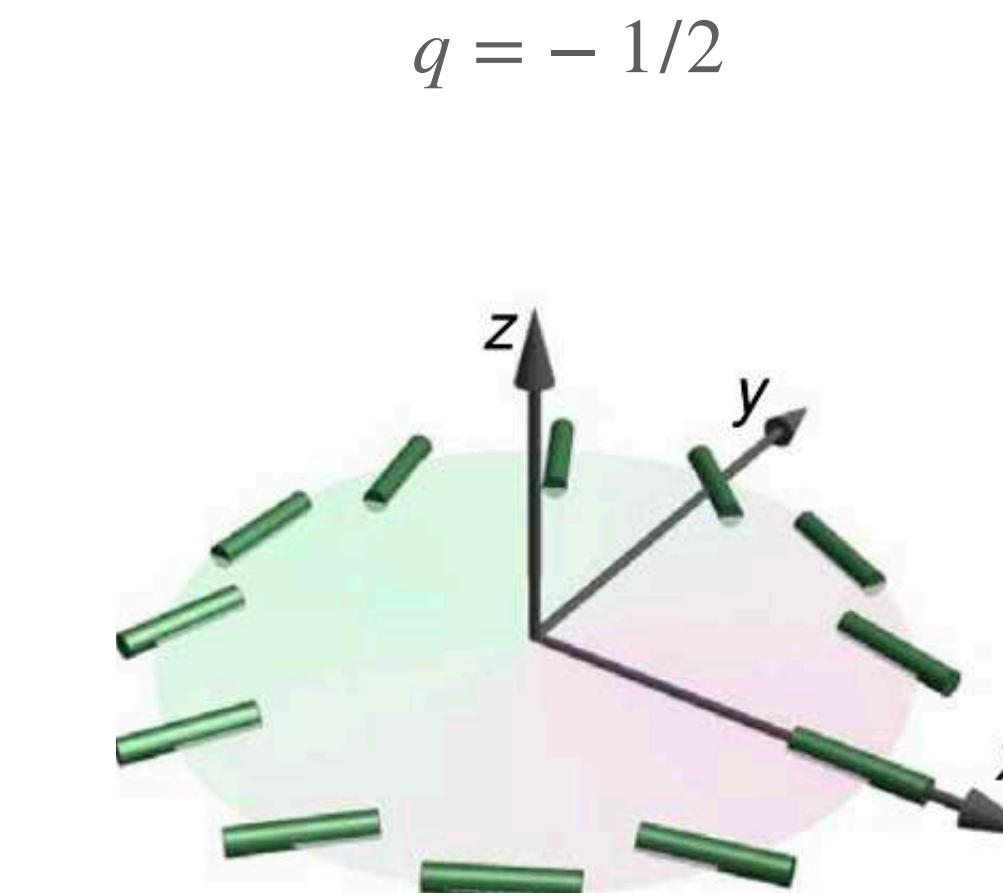


Uni-axial nematic LC

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



$q = +1/2$



$q = -1/2$

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

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

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

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

Unit-norm quaternion

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

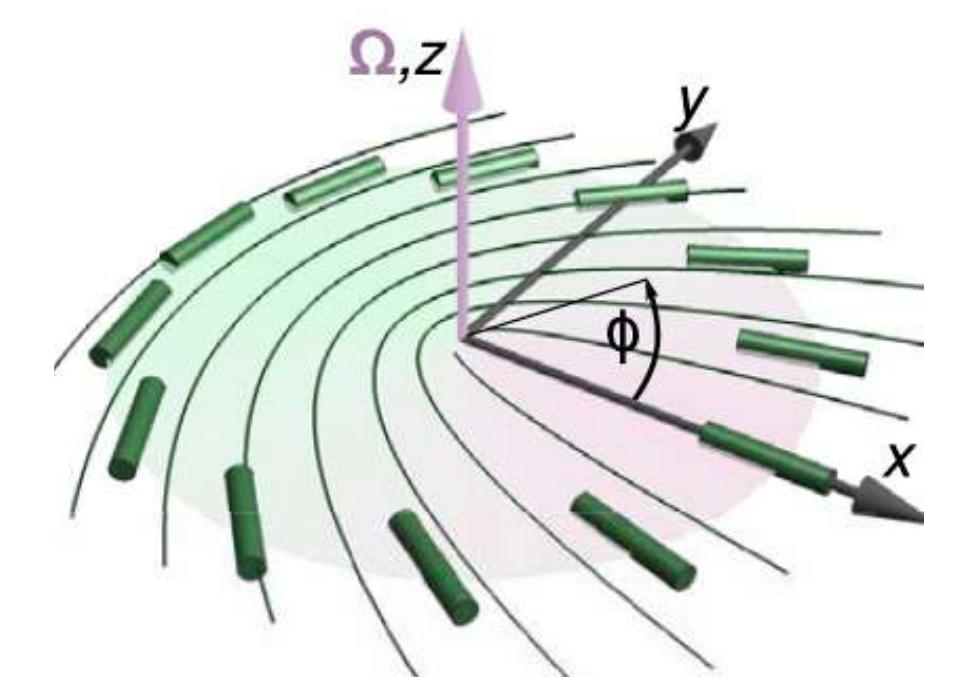
↔

Unit-norm quaternions equivalent to SU(2)

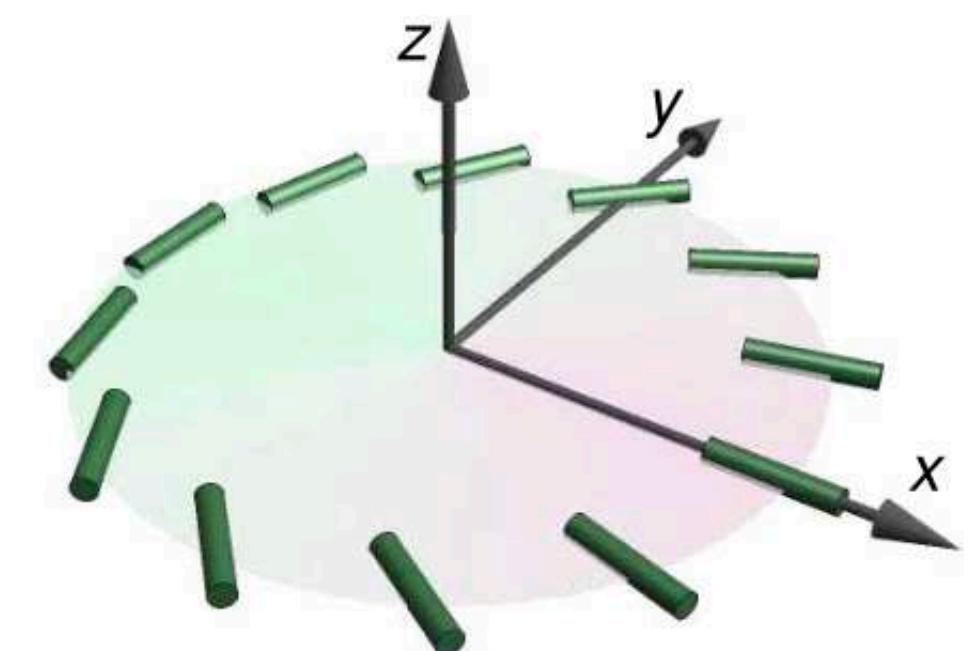
$$\begin{pmatrix} c_1 & -c_2^* \\ c_2 & c_1^* \end{pmatrix} \quad |c_1|^2 + |c_2|^2 = 1$$

Fully specified by first column

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

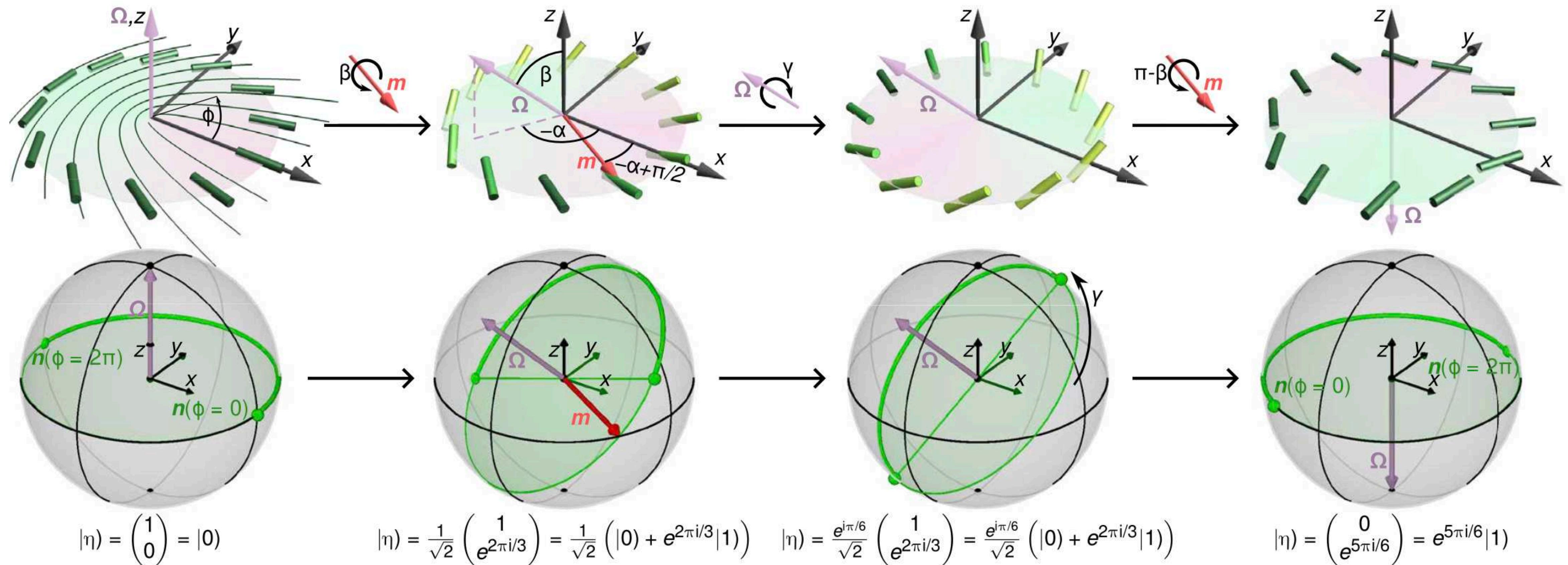


$$|0\rangle = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$



$$|1\rangle = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

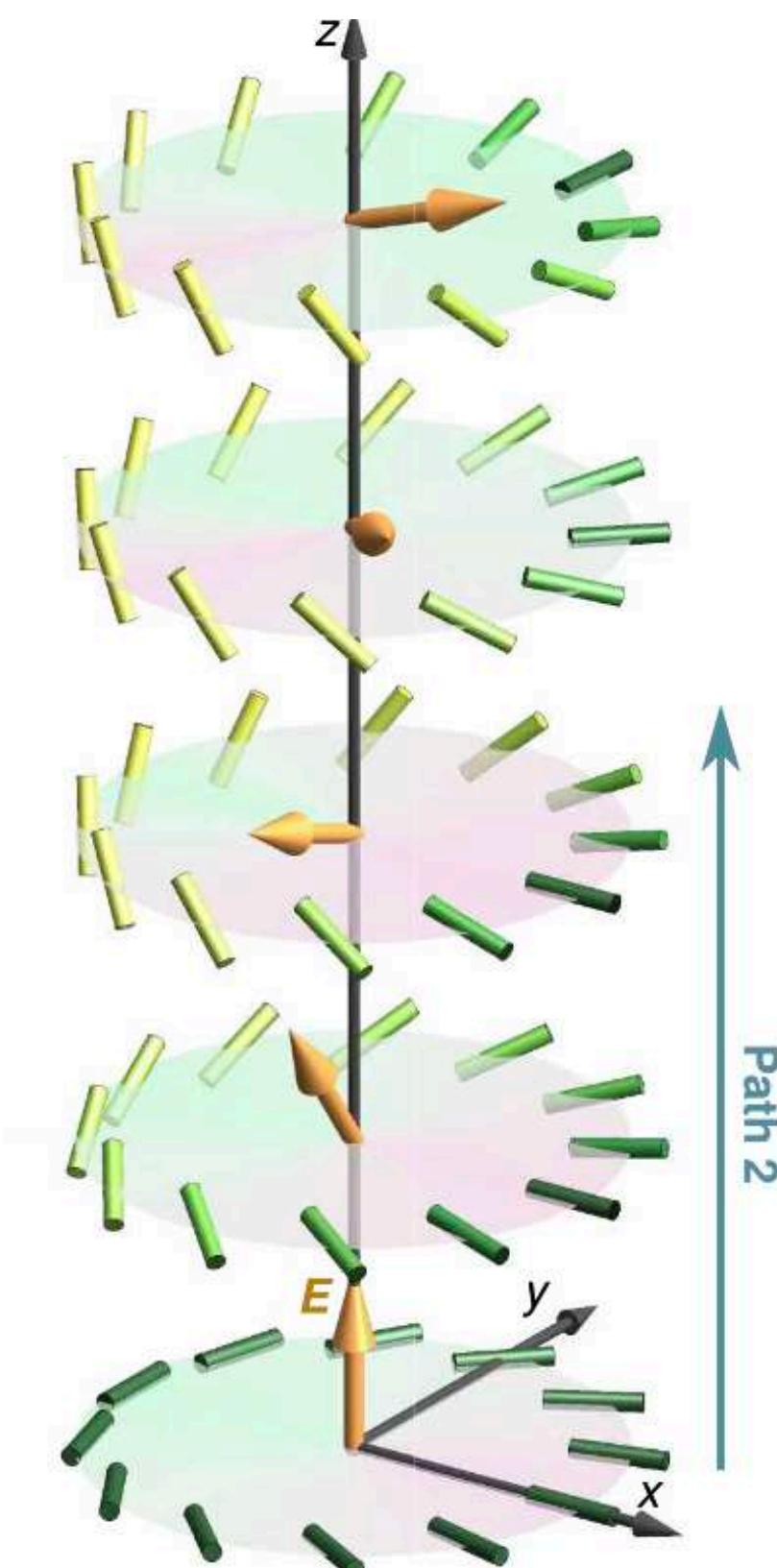
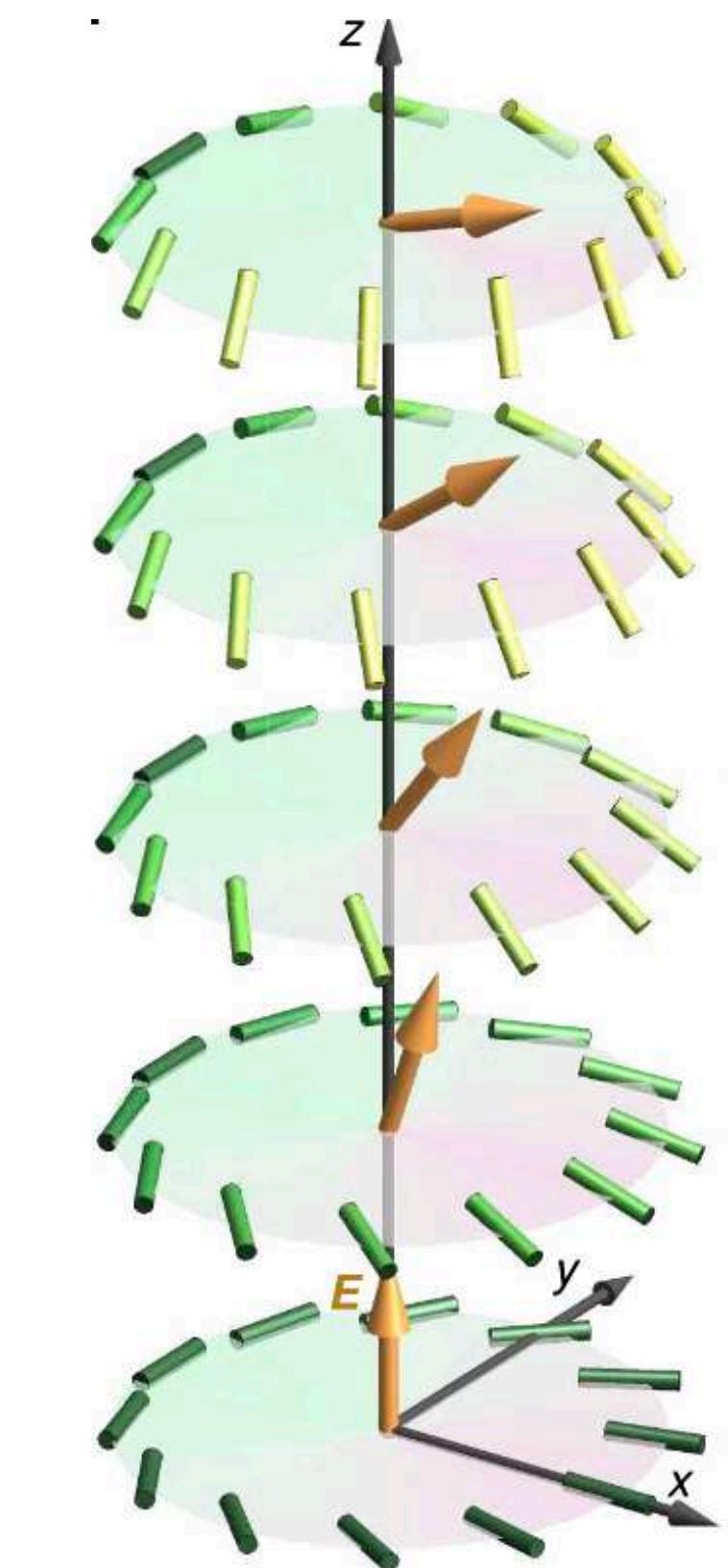
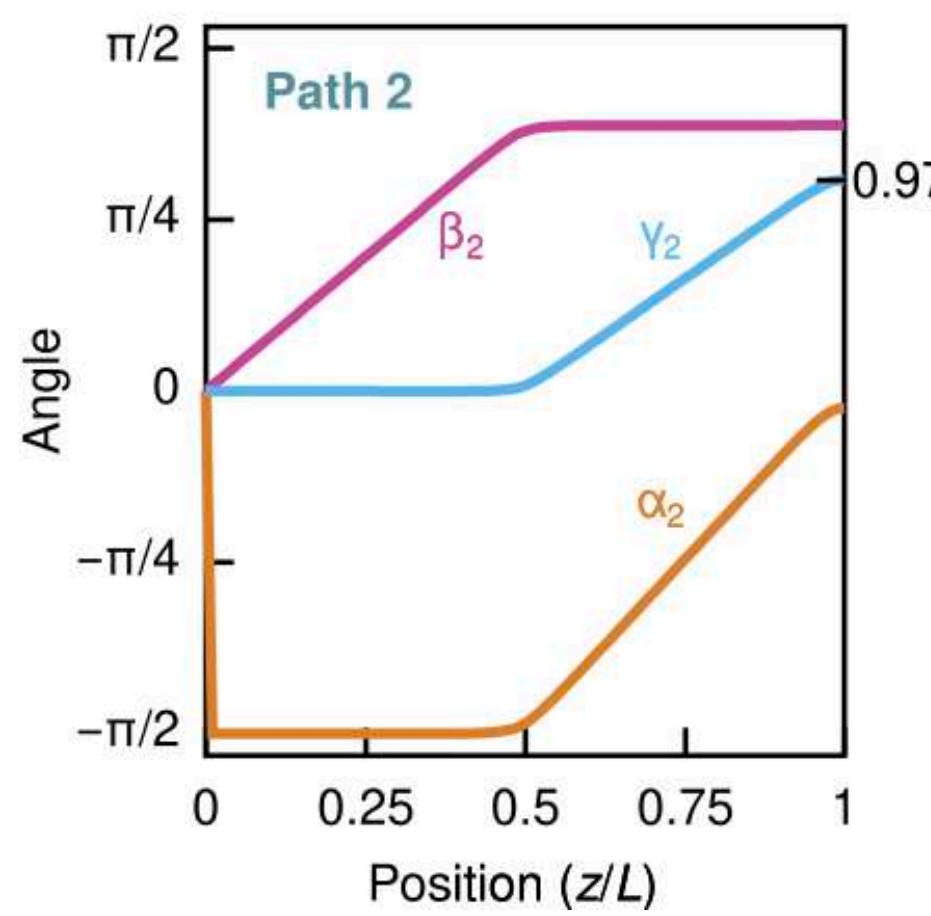
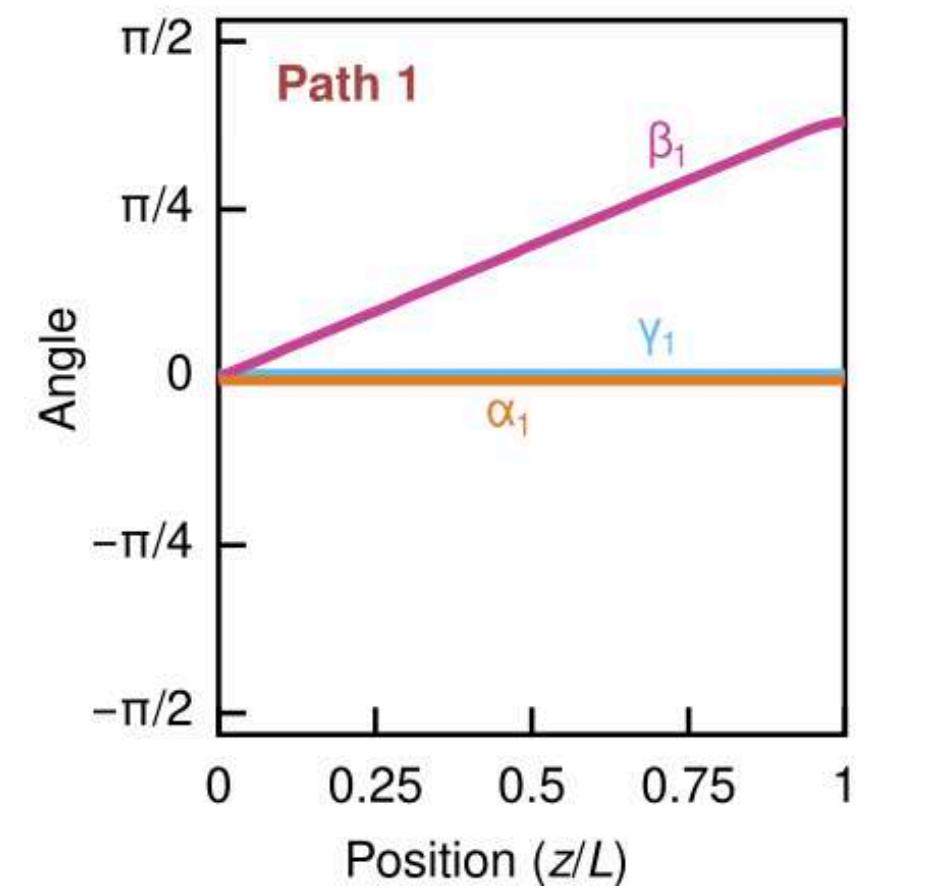
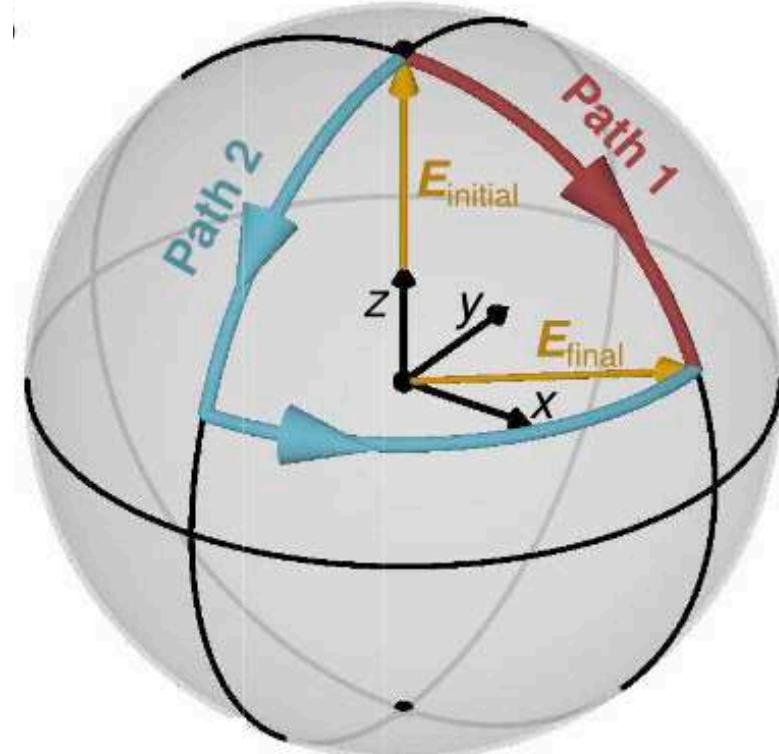
$$|\eta\rangle = c_1|0\rangle + c_2|1\rangle = e^{i\frac{\gamma}{2}} \left[\cos \frac{\beta}{2} |0\rangle + e^{-i\alpha} \sin \frac{\beta}{2} |1\rangle \right]$$



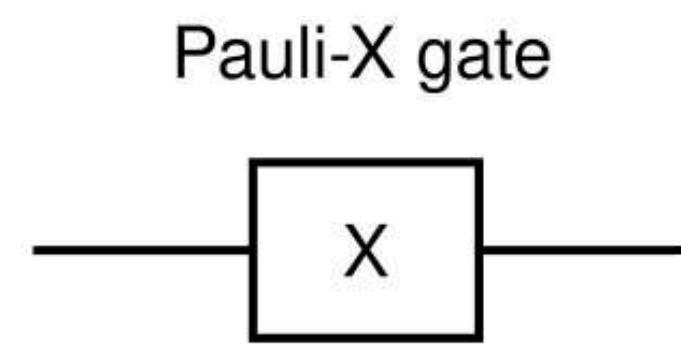
nbit manipulations with electric fields



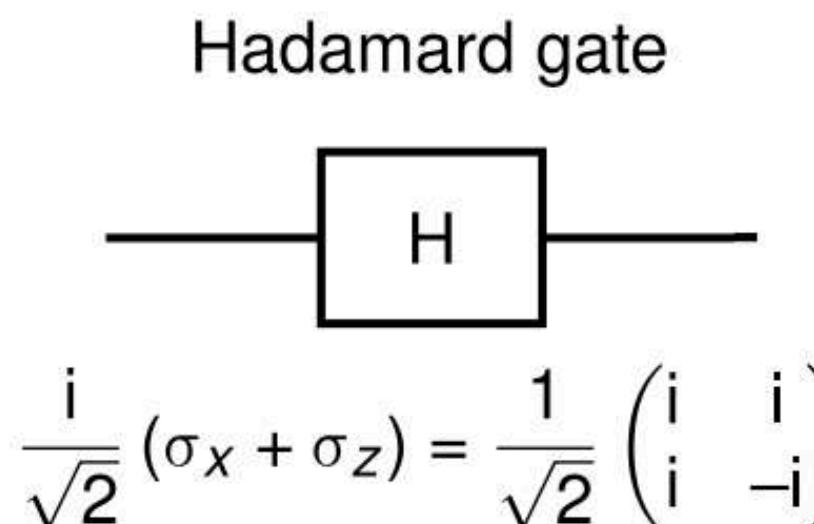
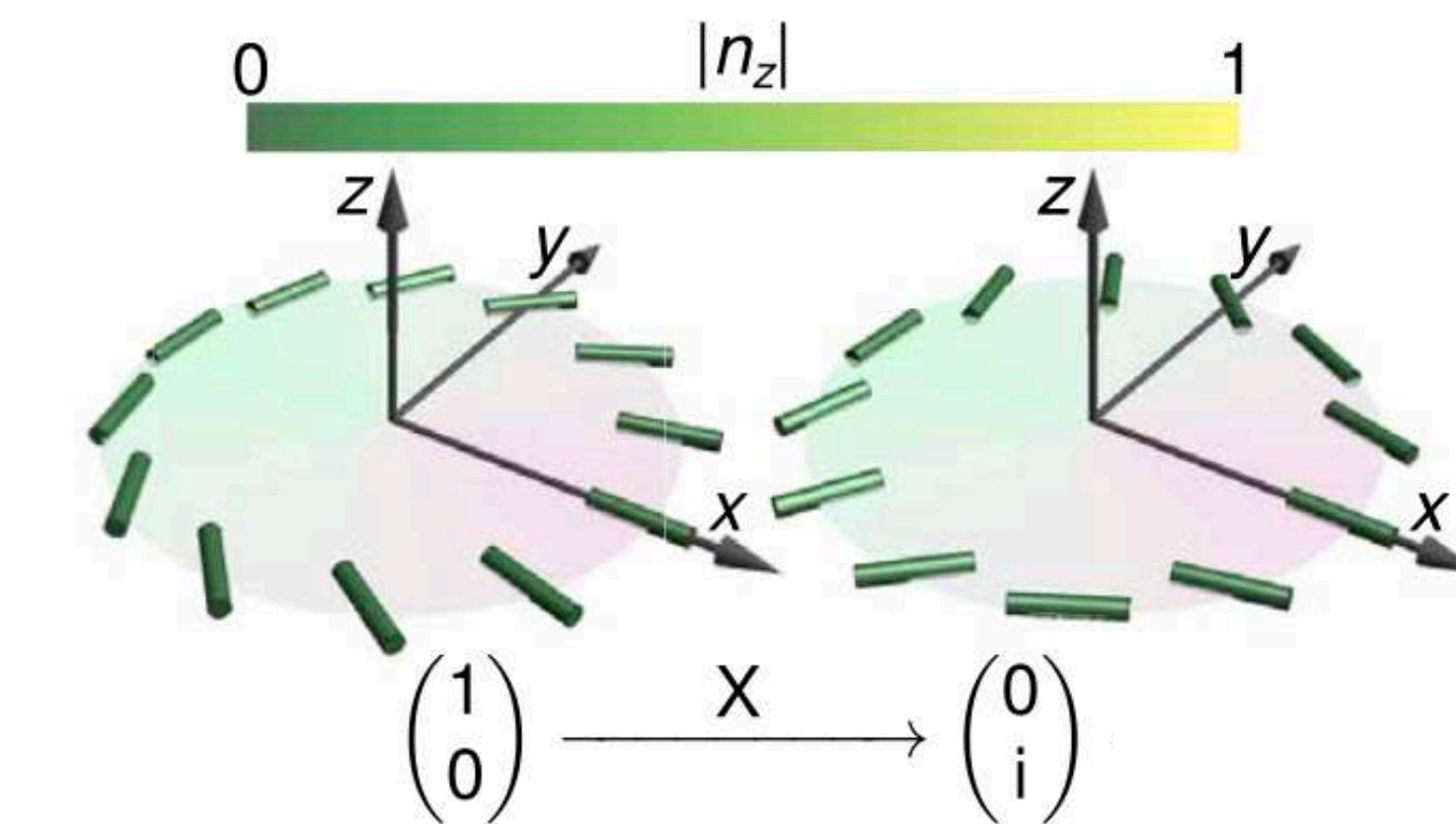
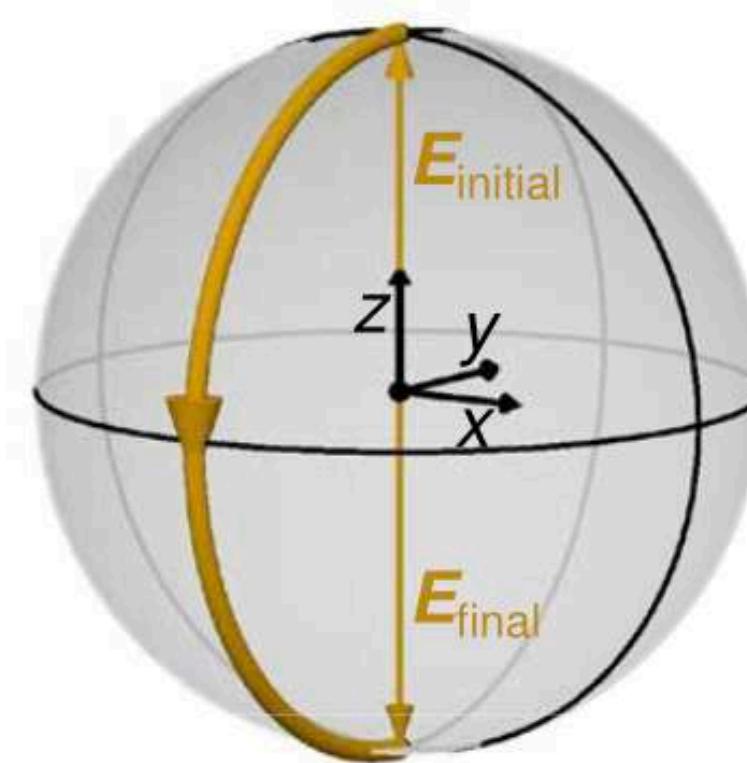
$$\frac{d\eta}{dt} = \frac{K}{\Gamma} \frac{\partial^2 \eta}{\partial z^2} - \frac{\epsilon_a}{4\Gamma} \text{Tr} (E \eta \sigma_z \eta^\dagger) E \eta \sigma_z - \tilde{\lambda} \eta, \quad \tilde{\lambda} = \frac{\lambda}{\pi \Gamma}$$



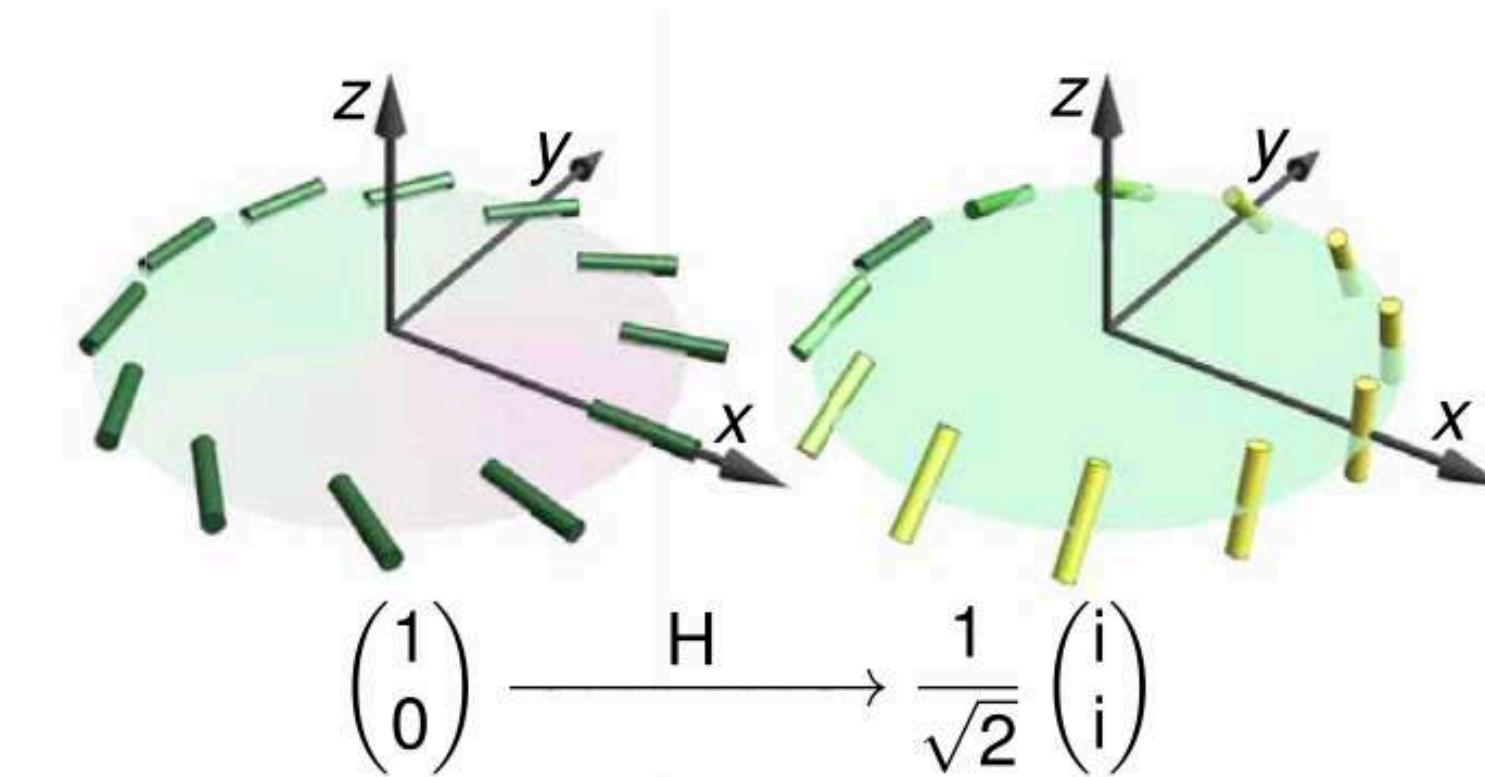
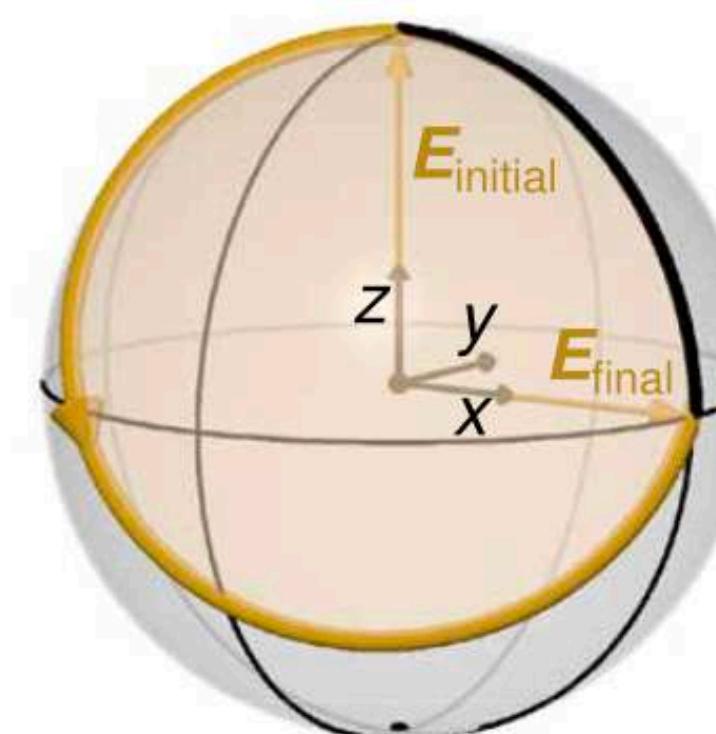
single nbit-operations



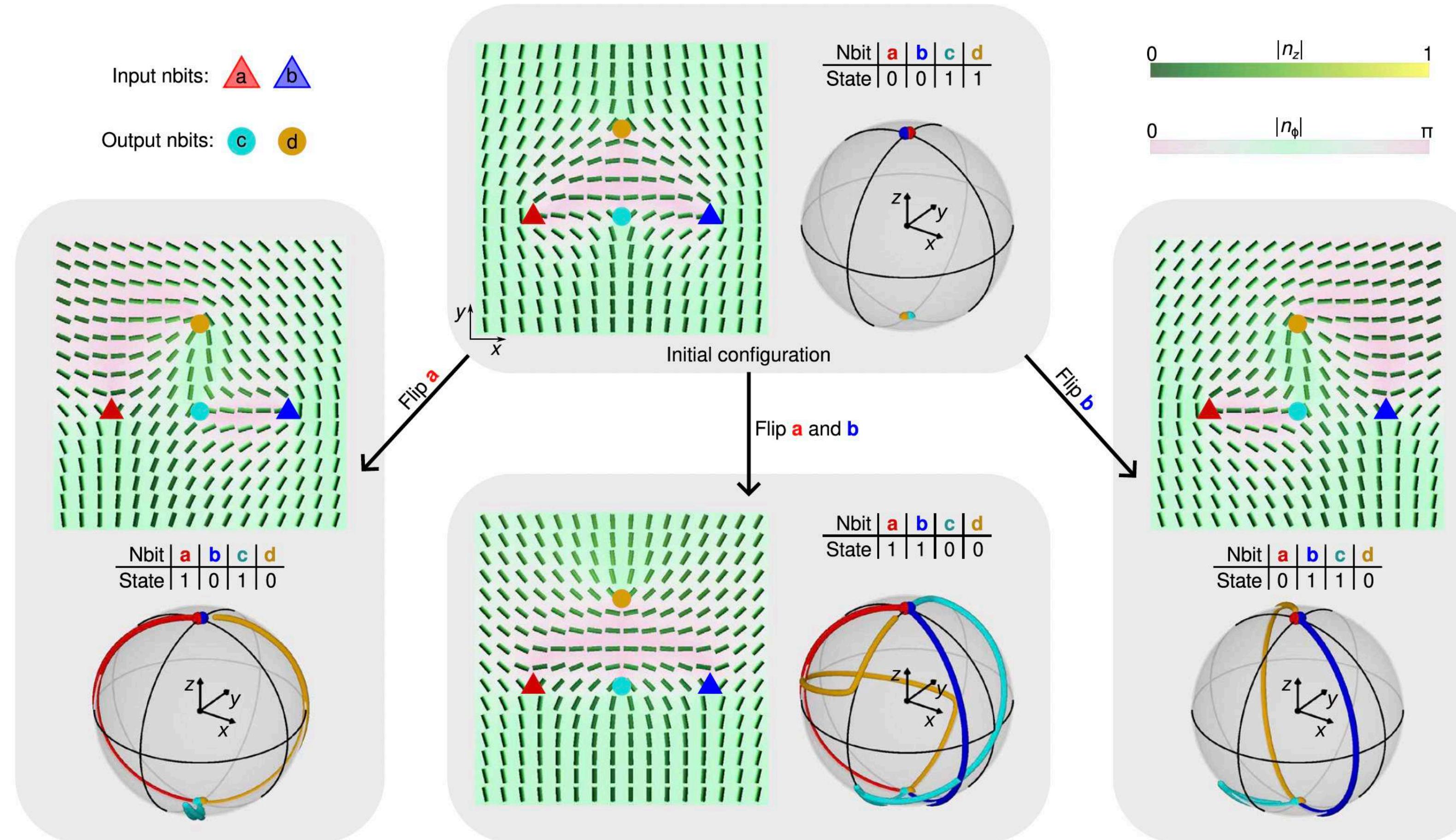
$$i\sigma_X = \begin{pmatrix} 0 & i \\ i & 0 \end{pmatrix}$$



$$\frac{i}{\sqrt{2}} (\sigma_X + \sigma_Z) = \frac{1}{\sqrt{2}} \begin{pmatrix} i & i \\ i & -i \end{pmatrix}$$



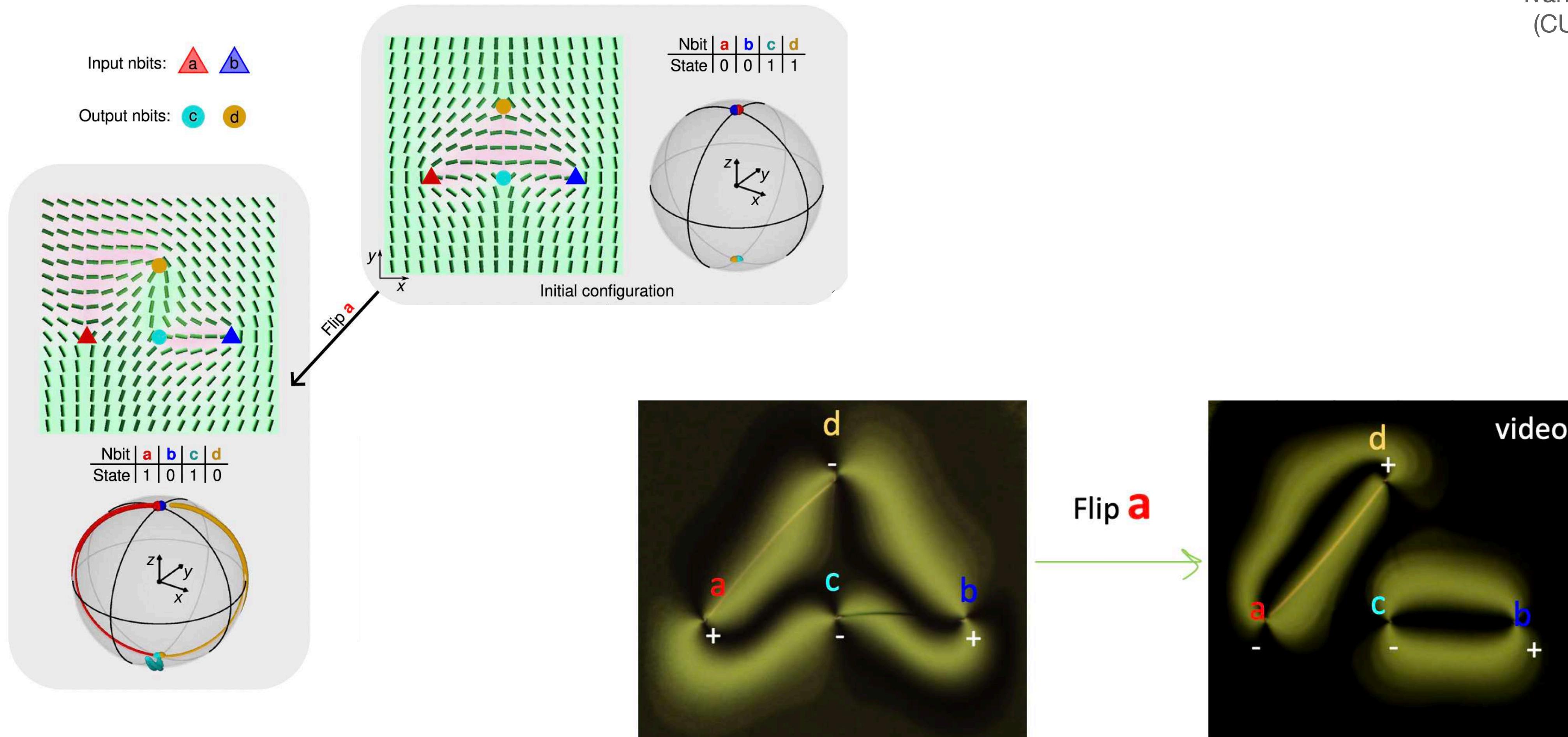
Classical nbit-logic : universal gates



Recent experiments: towards nbit logic gates



Ivan Smalyukh
(CU Boulder)



Recent experiments: towards nbit memory



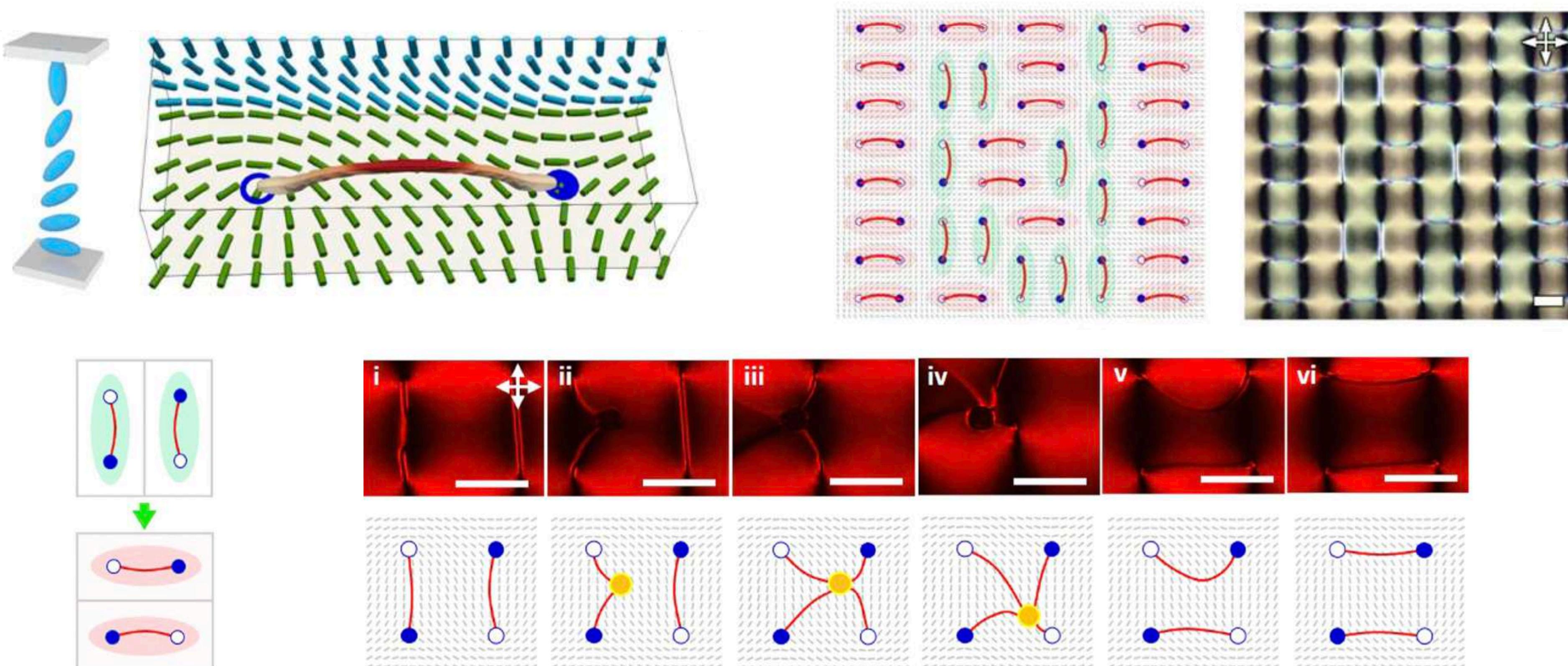
PHYSICAL REVIEW X 15, 021084 (2025)

Featured in Physics

Ivan Smalyukh
(CU Boulder)

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

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

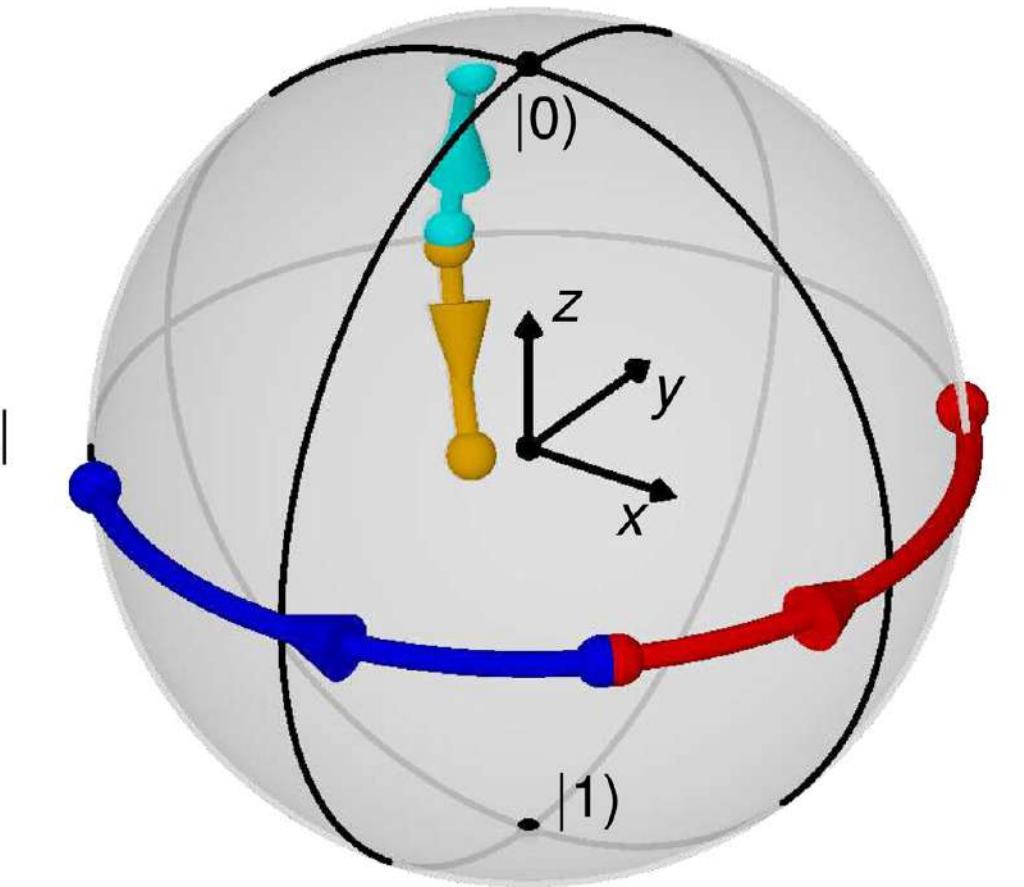
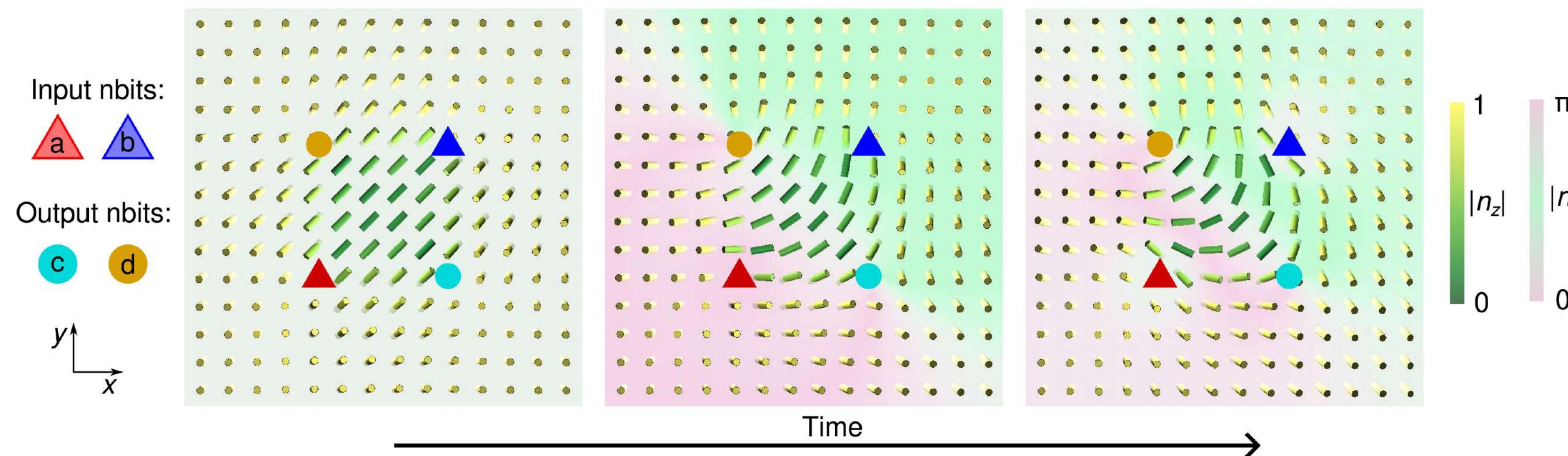


Outlook: beyond classical bit logic



Ziga Kos

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



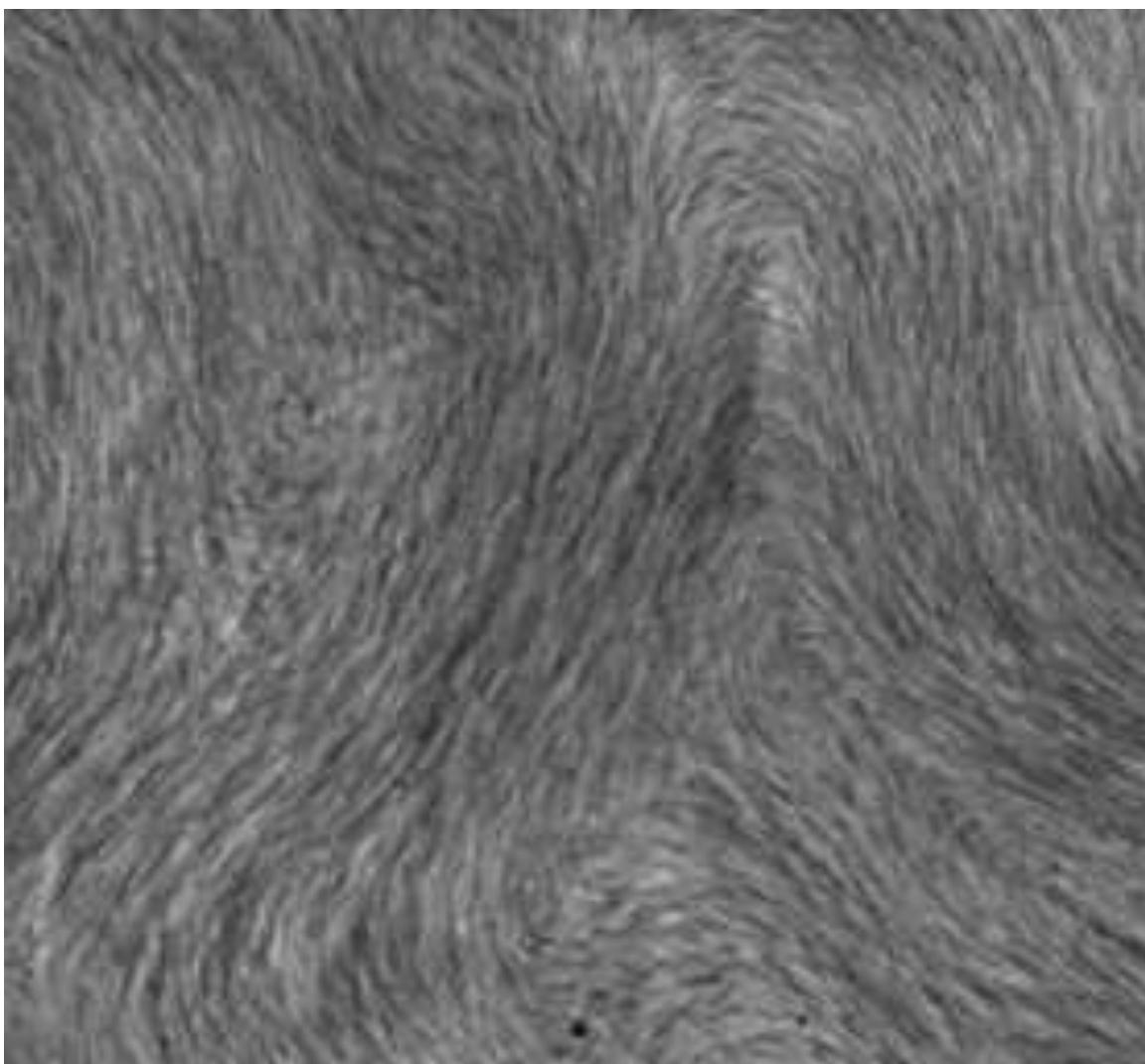
What is the space of controllable non-classical bit gate operations ?



Petur Bryde

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

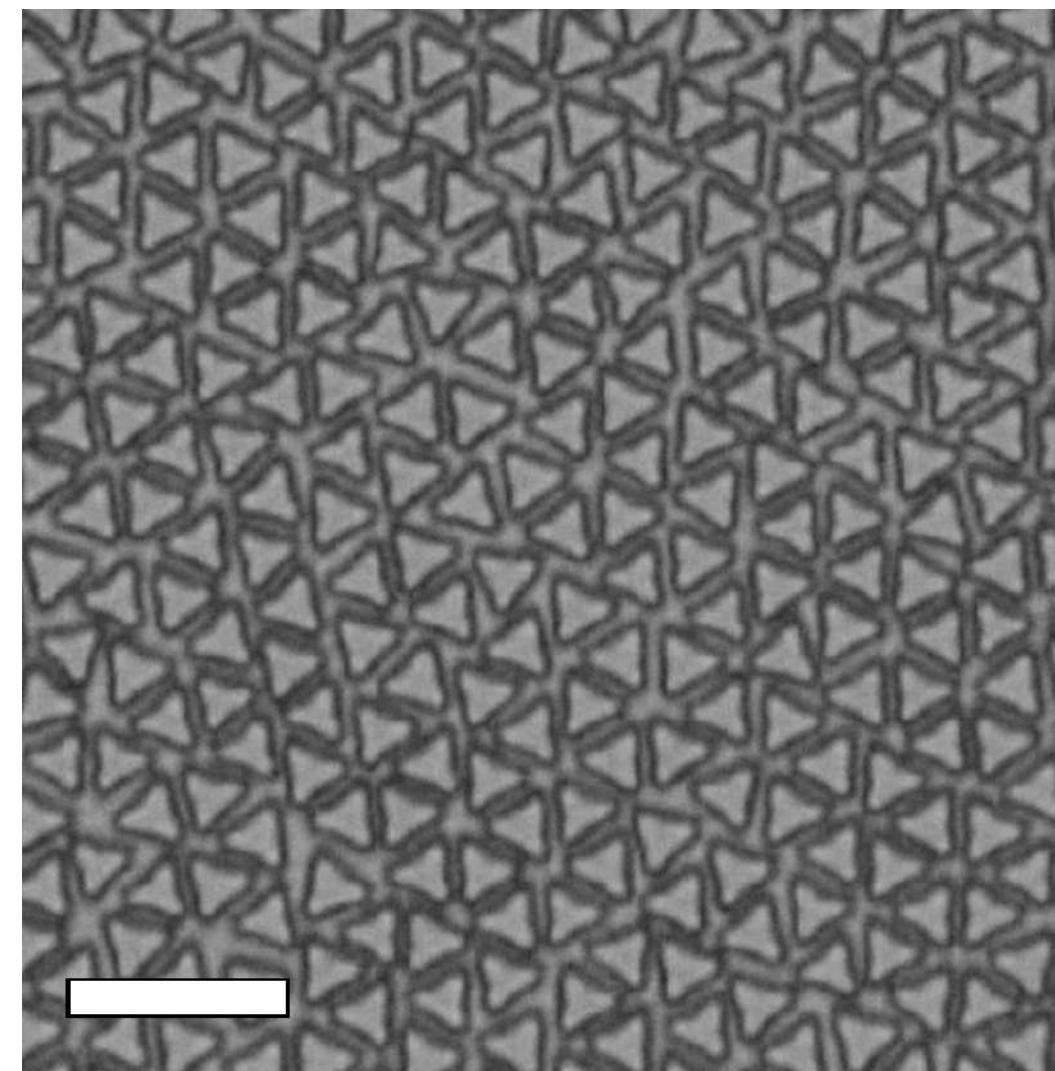
2-atic
(nematic)



Zhou et al. PNAS 2014

bacteria

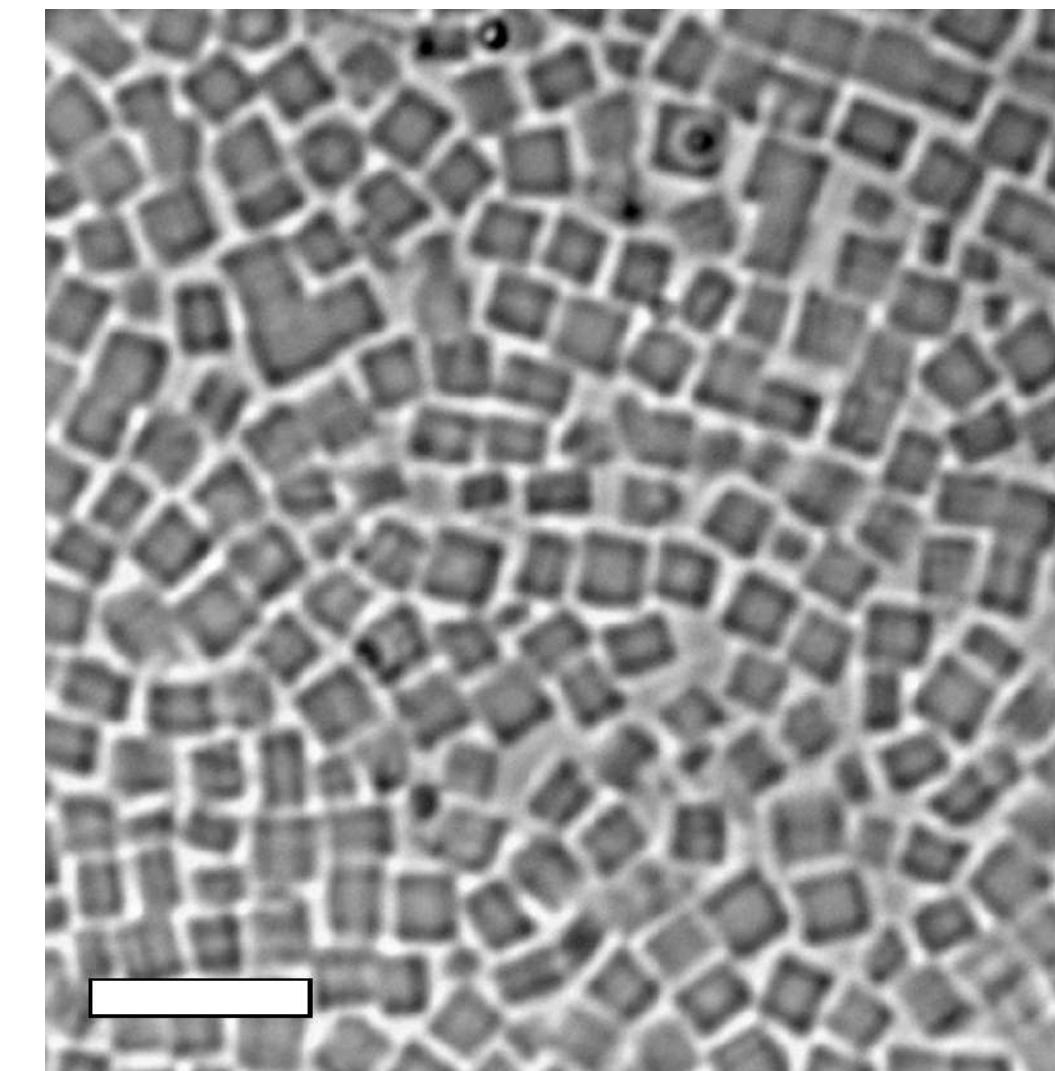
3-atic



Zhao et al. Nat Comm 2012

colloids

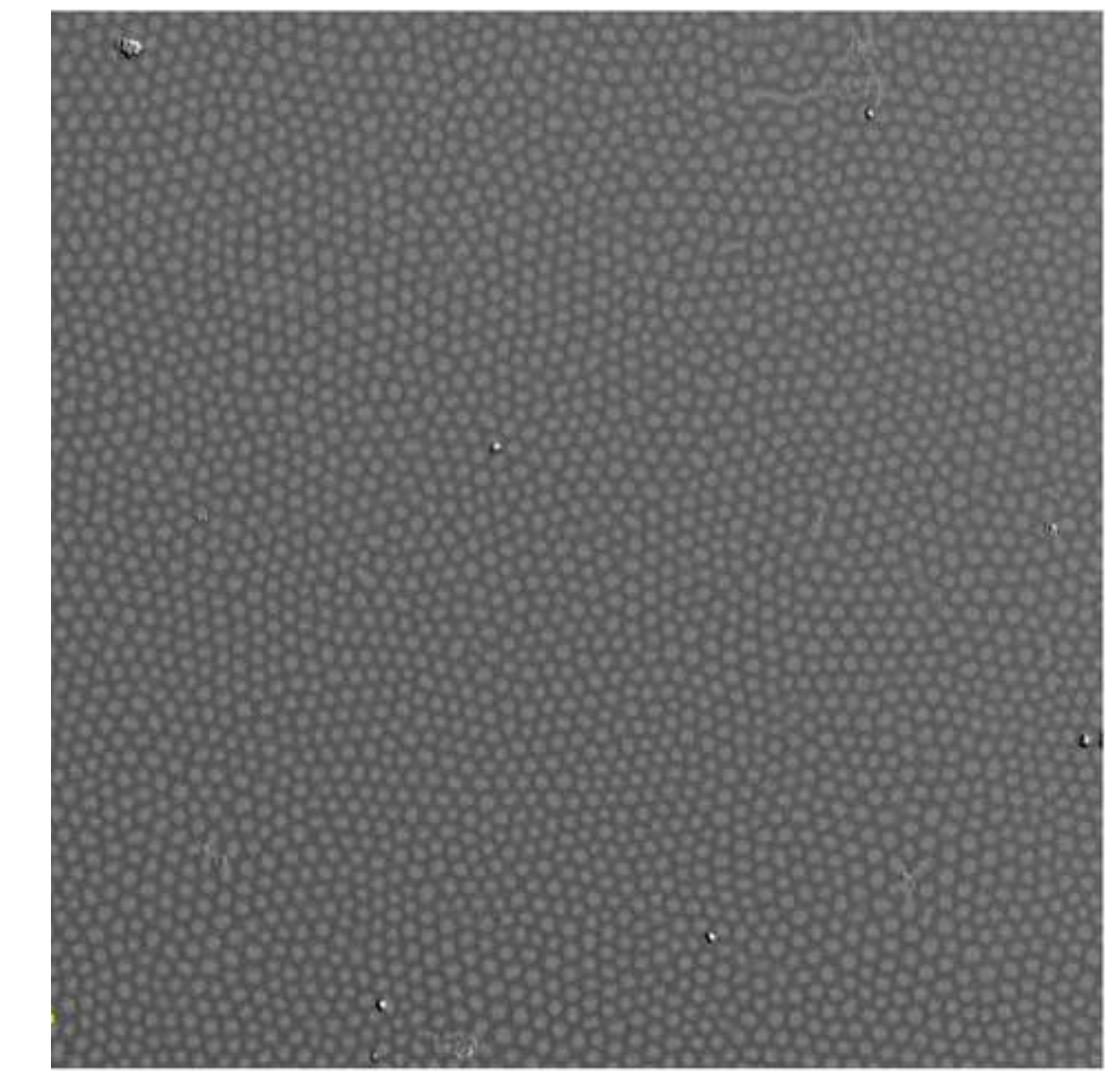
4-atic



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

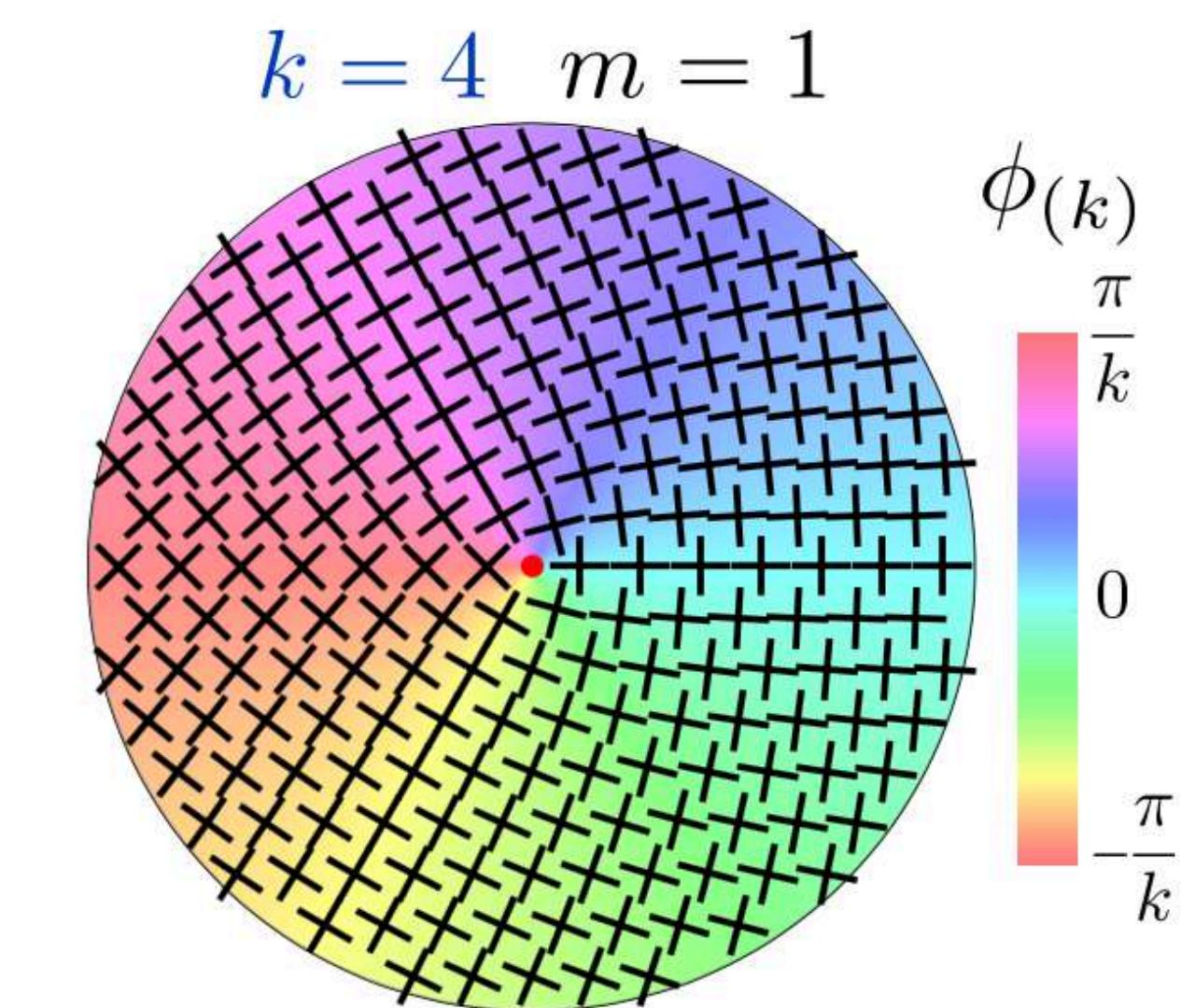
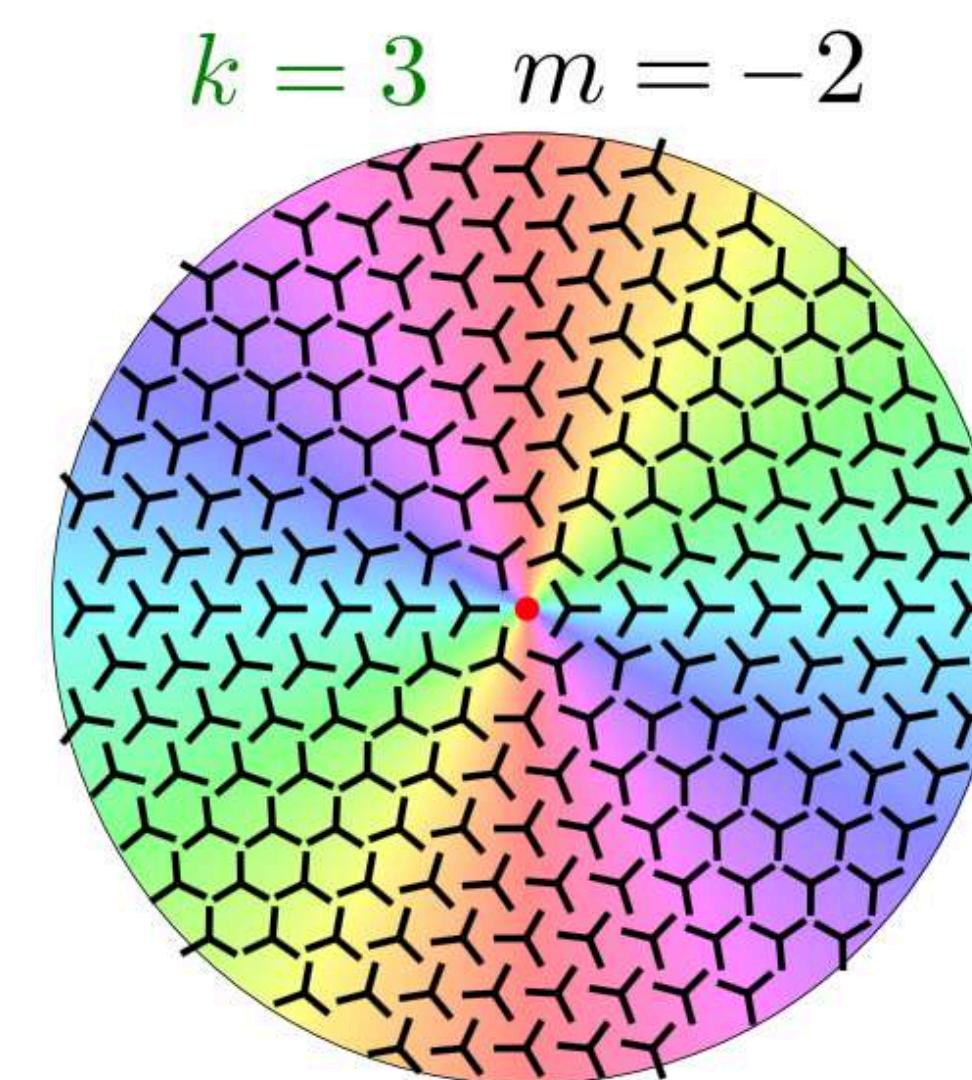
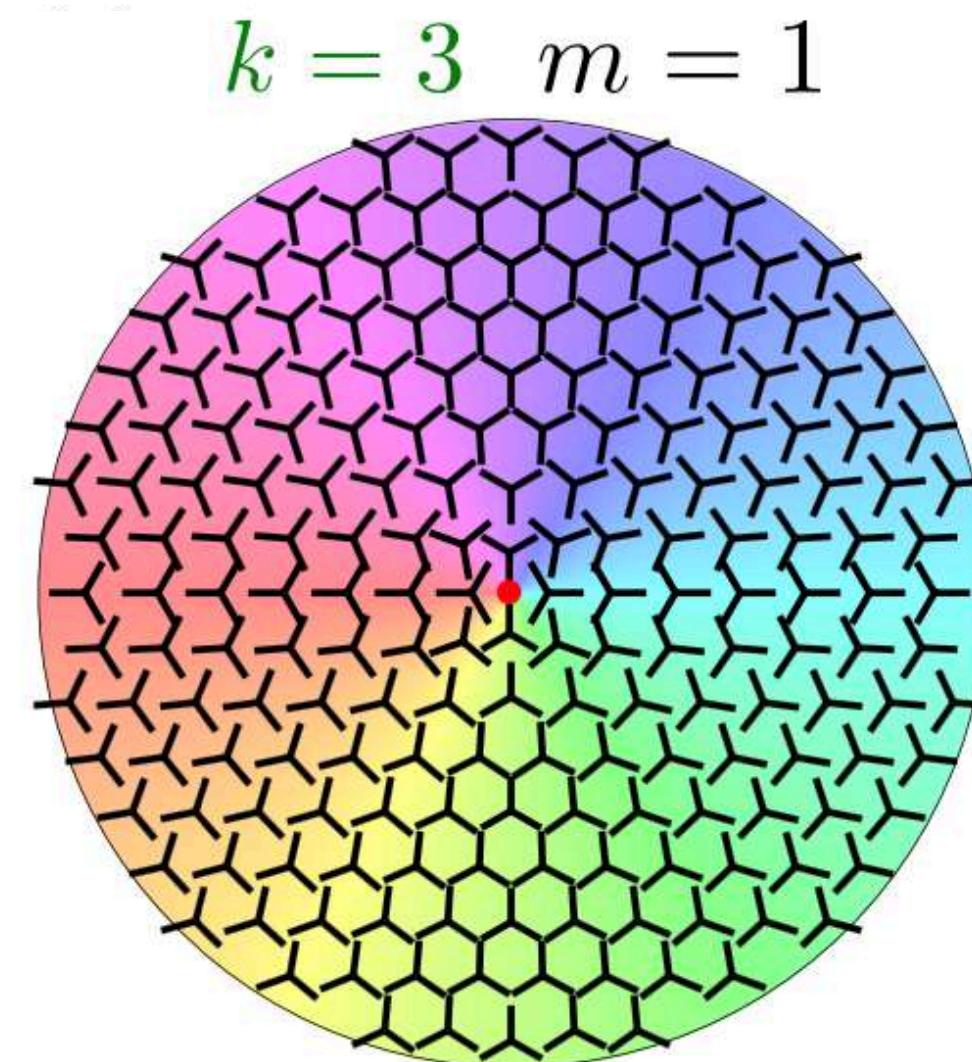
Fractional defects in k -atic LCs

Complex k -atic order parameter

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

Topological charge

$$q = \frac{1}{2\pi} \oint_C d\mathbf{l} \cdot \nabla \phi_{(k)} = \frac{m}{k}$$



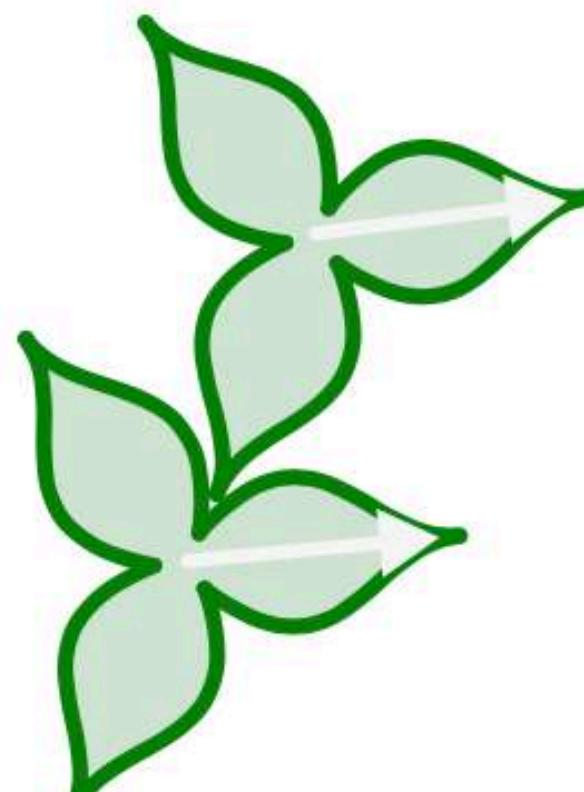
Minimal microscopic disordered-lattice model



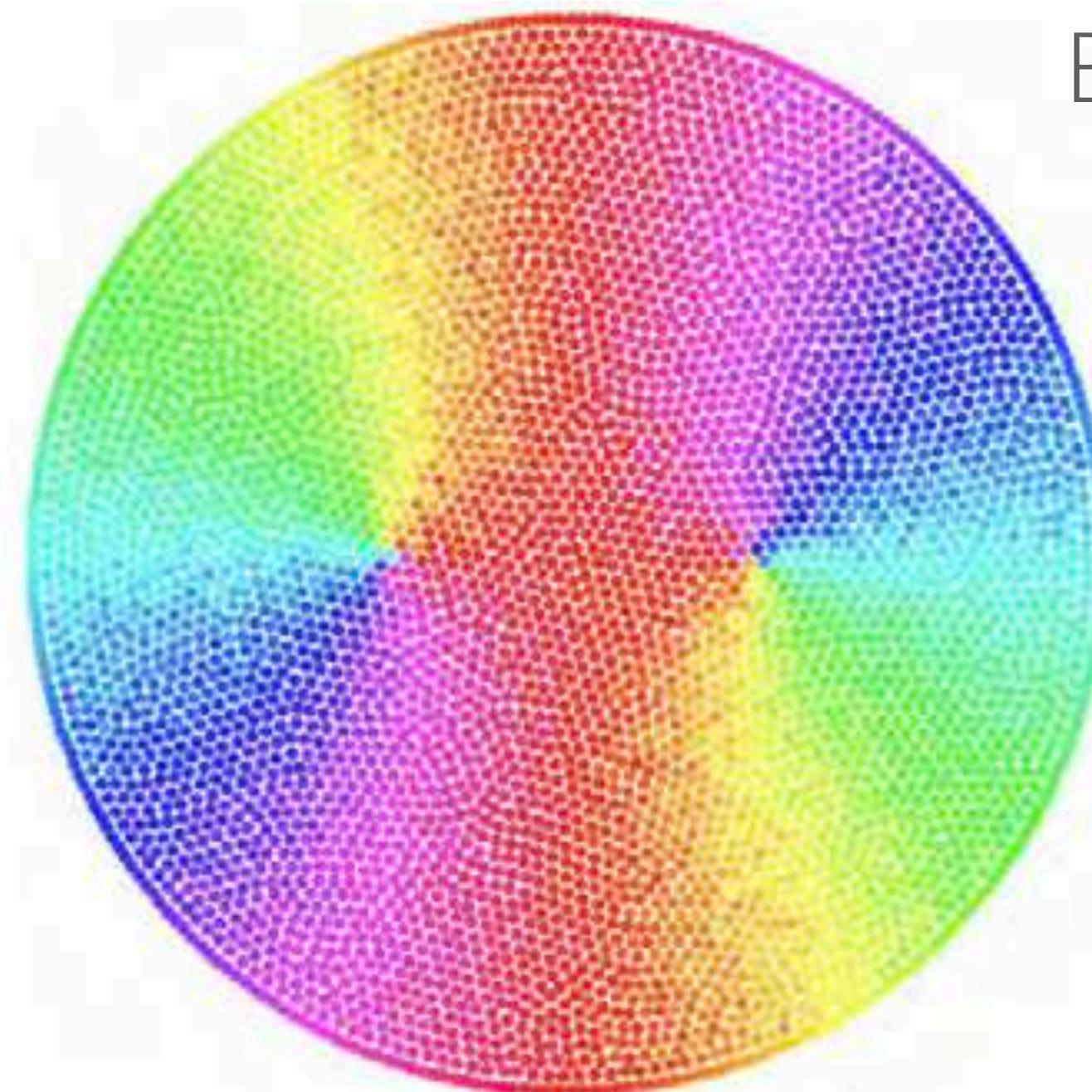
Alex Mietke

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

$g > 0$: alignment

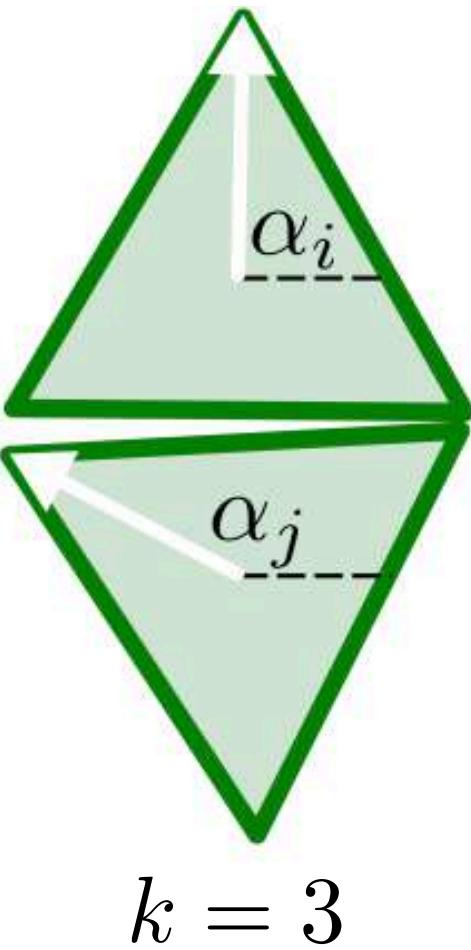
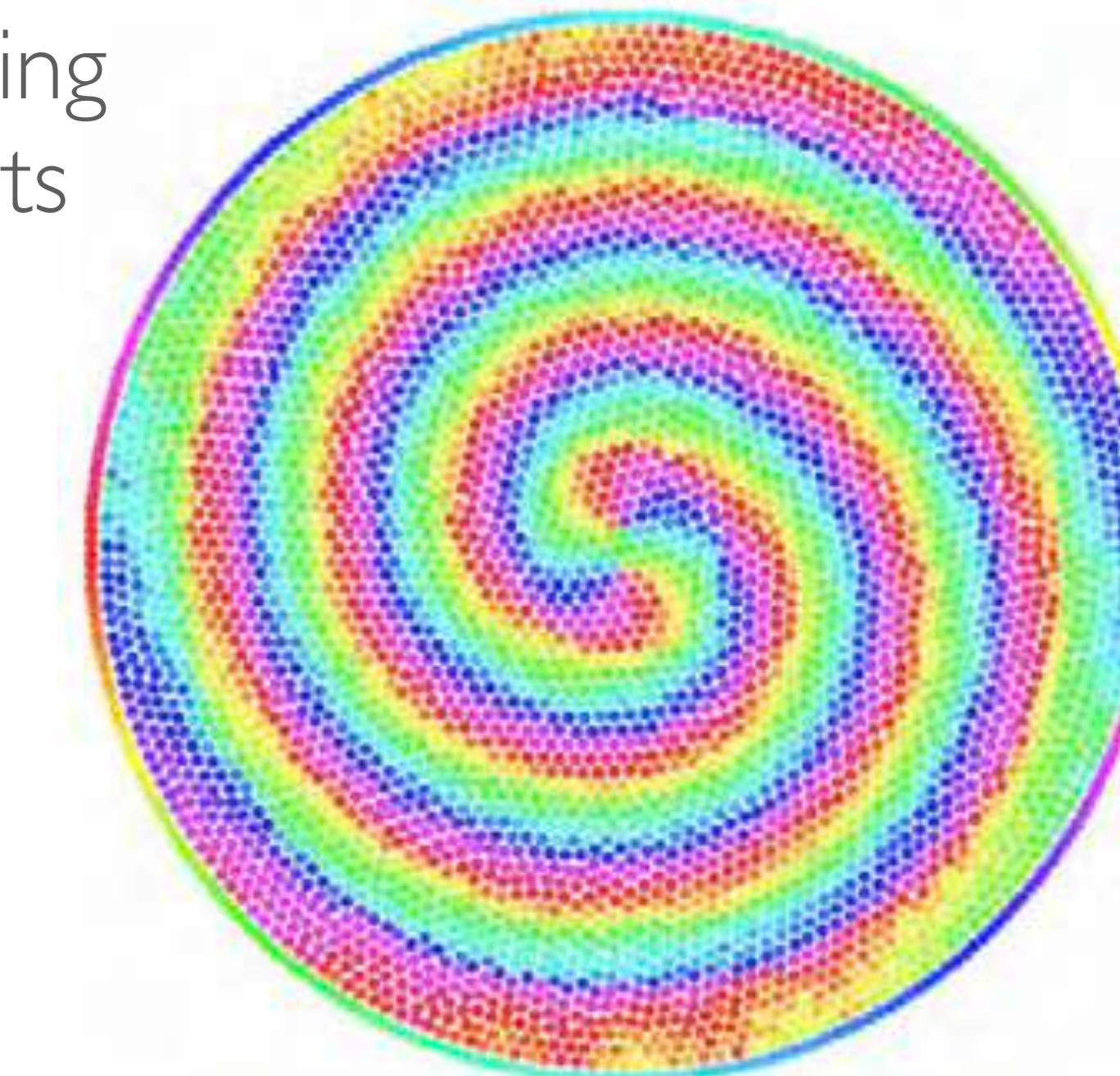


$k = 3$

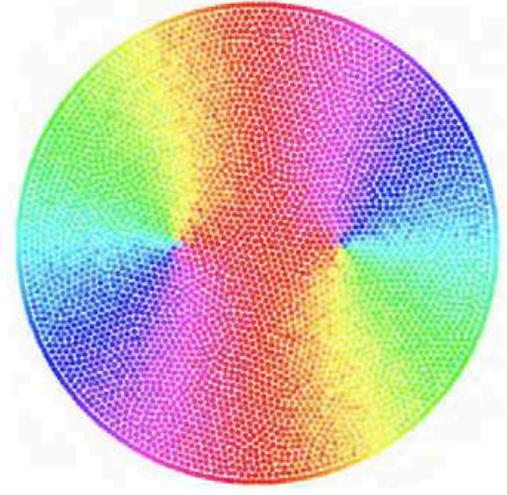


Boundary anchoring
enforces 2 defects

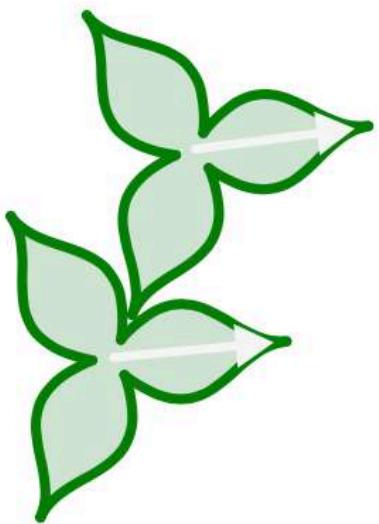
$g < 0$: anti-alignment



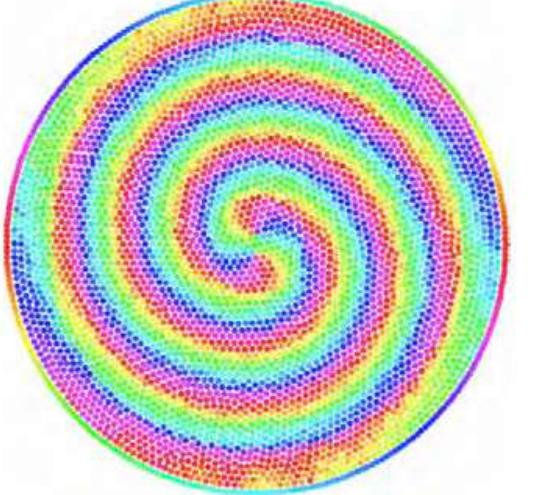
Minimal microscopic random-lattice model



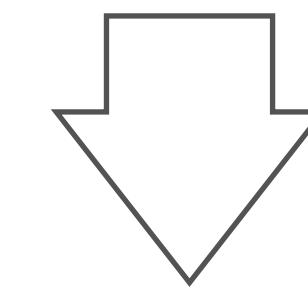
$g > 0$: alignment



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



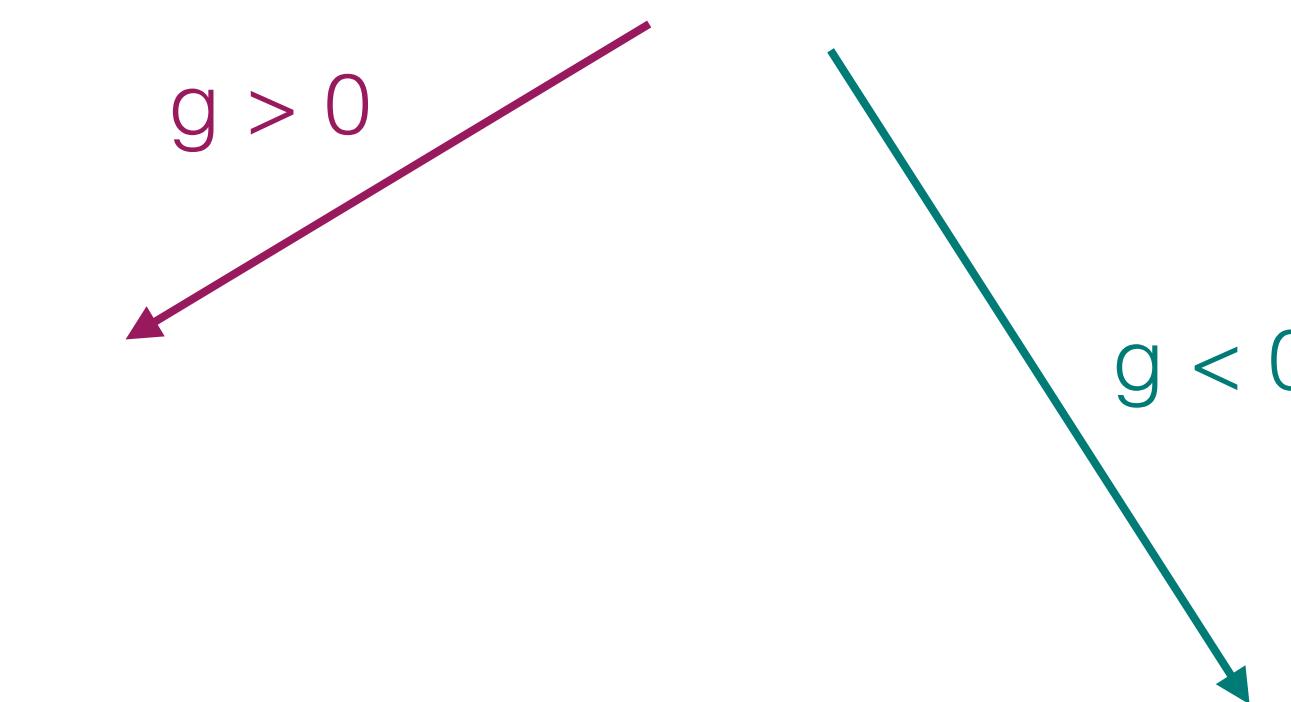
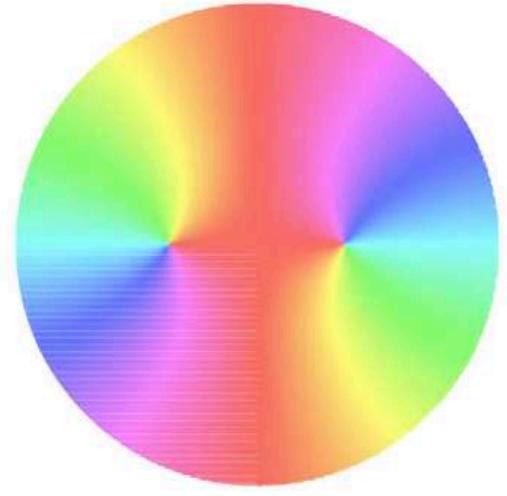
$g < 0$: anti-alignment



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

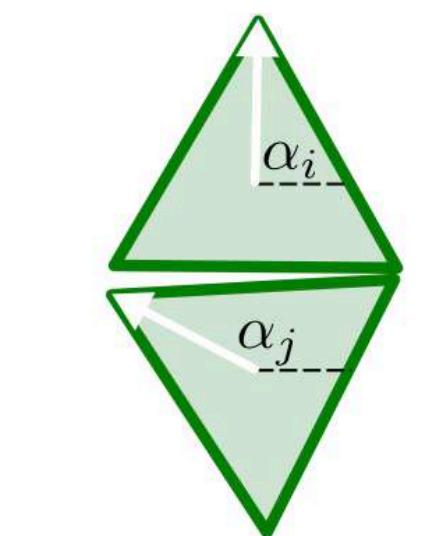
Ginzburg-Landau

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

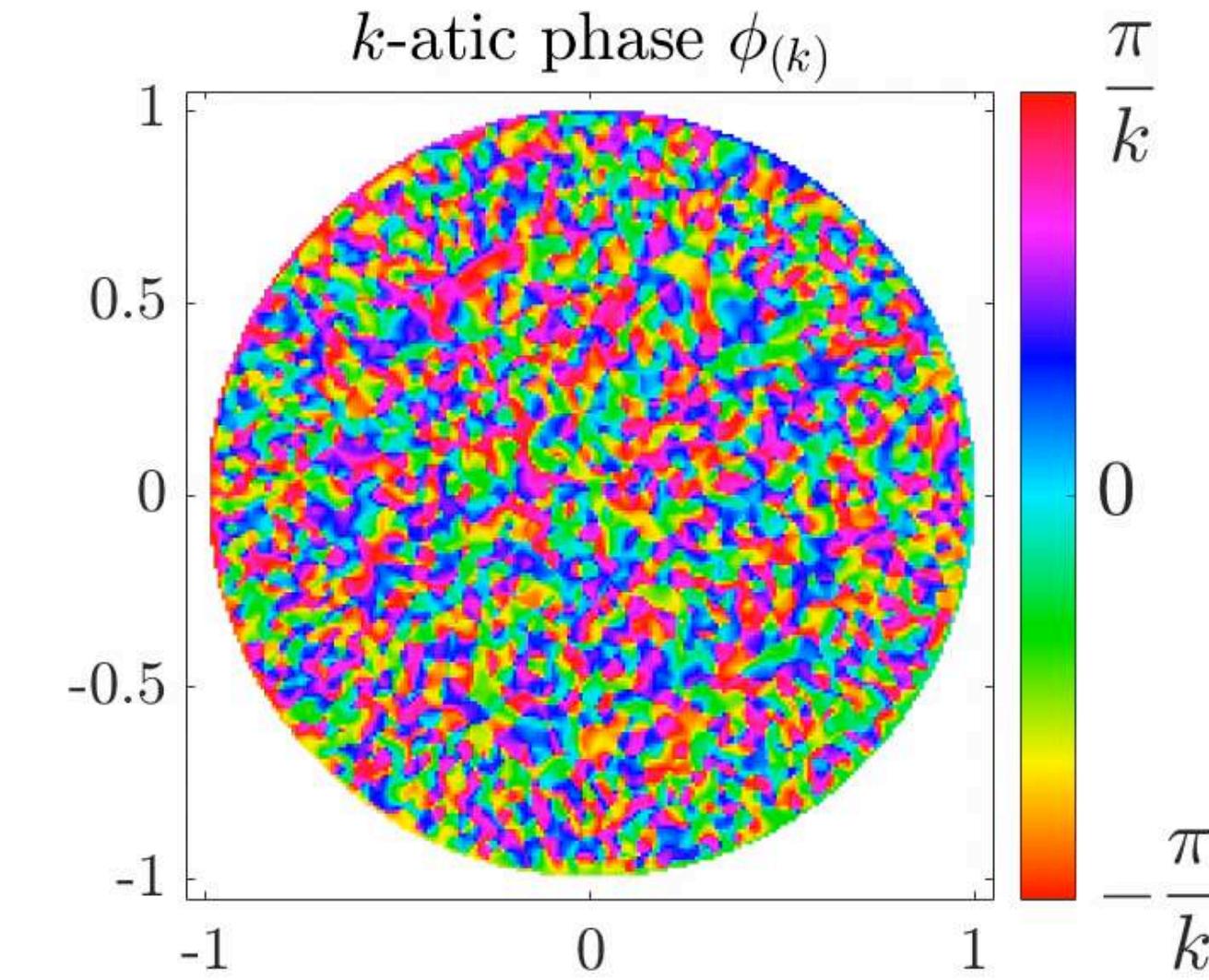
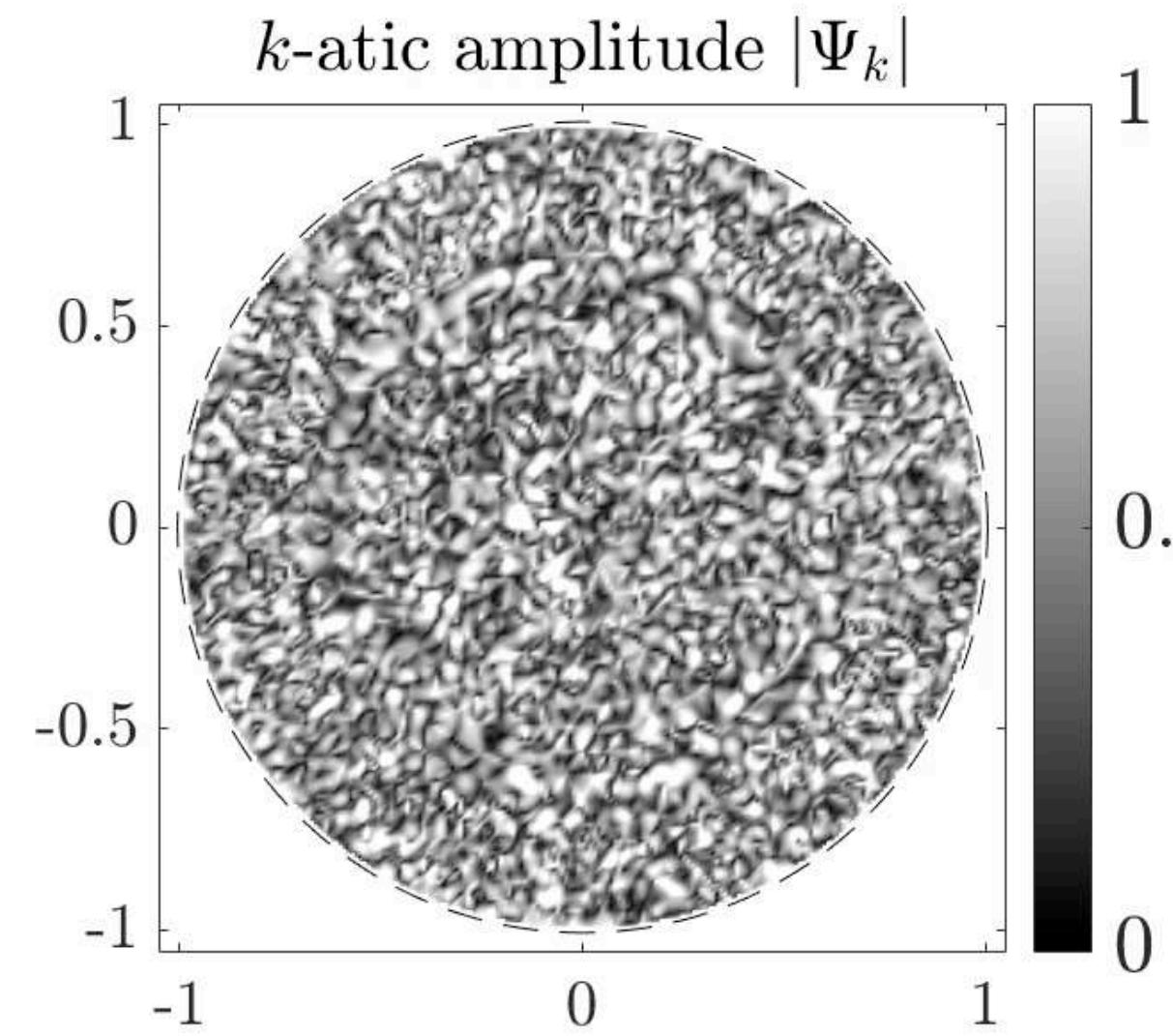


$$\mathcal{E}_k = \int dA (A |\Psi_k|^2 + B |\Psi_k|^4 - L_1^2 |\nabla \Psi_k|^2 + L_2^4 |\nabla^2 \Psi_k|^2)$$

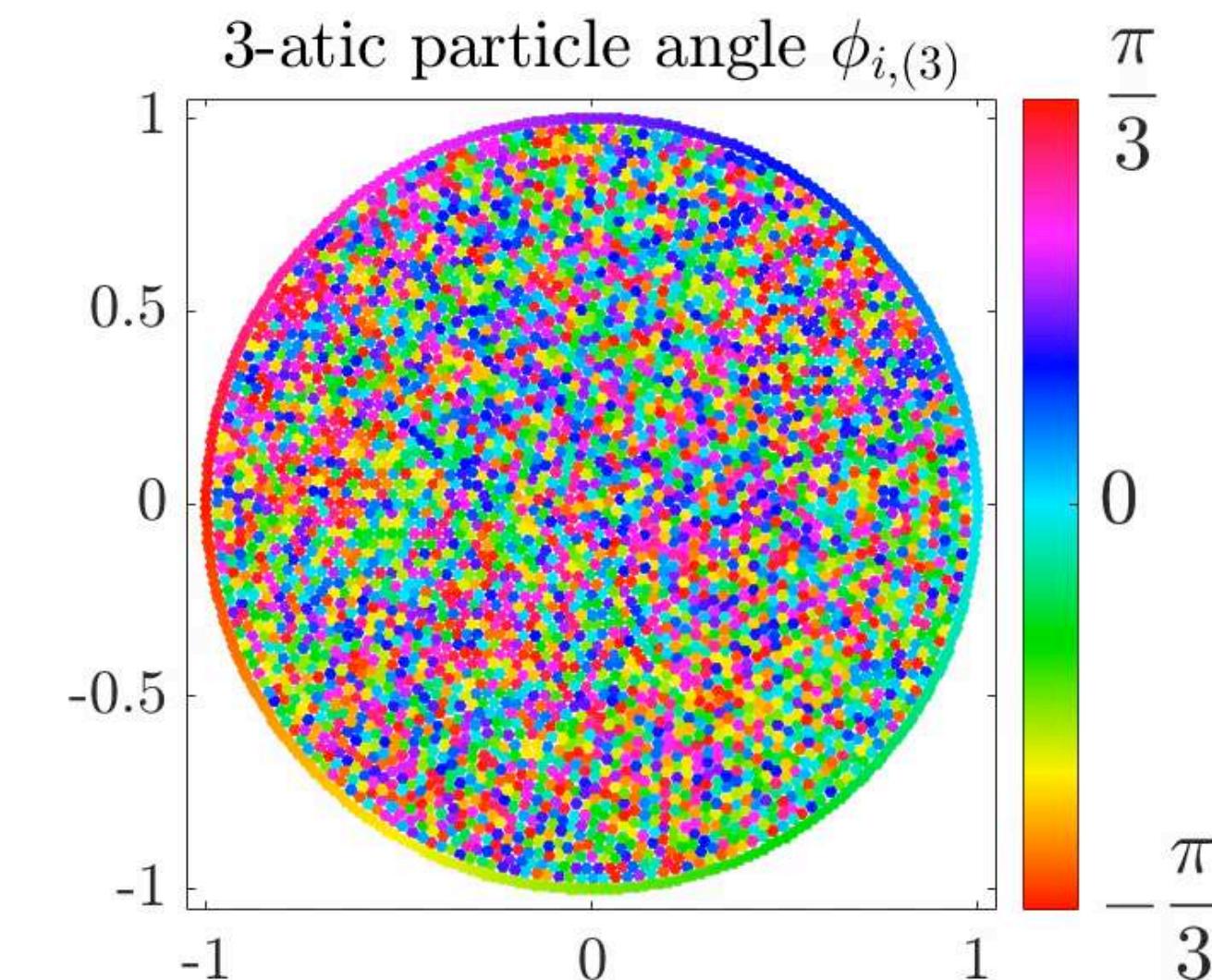
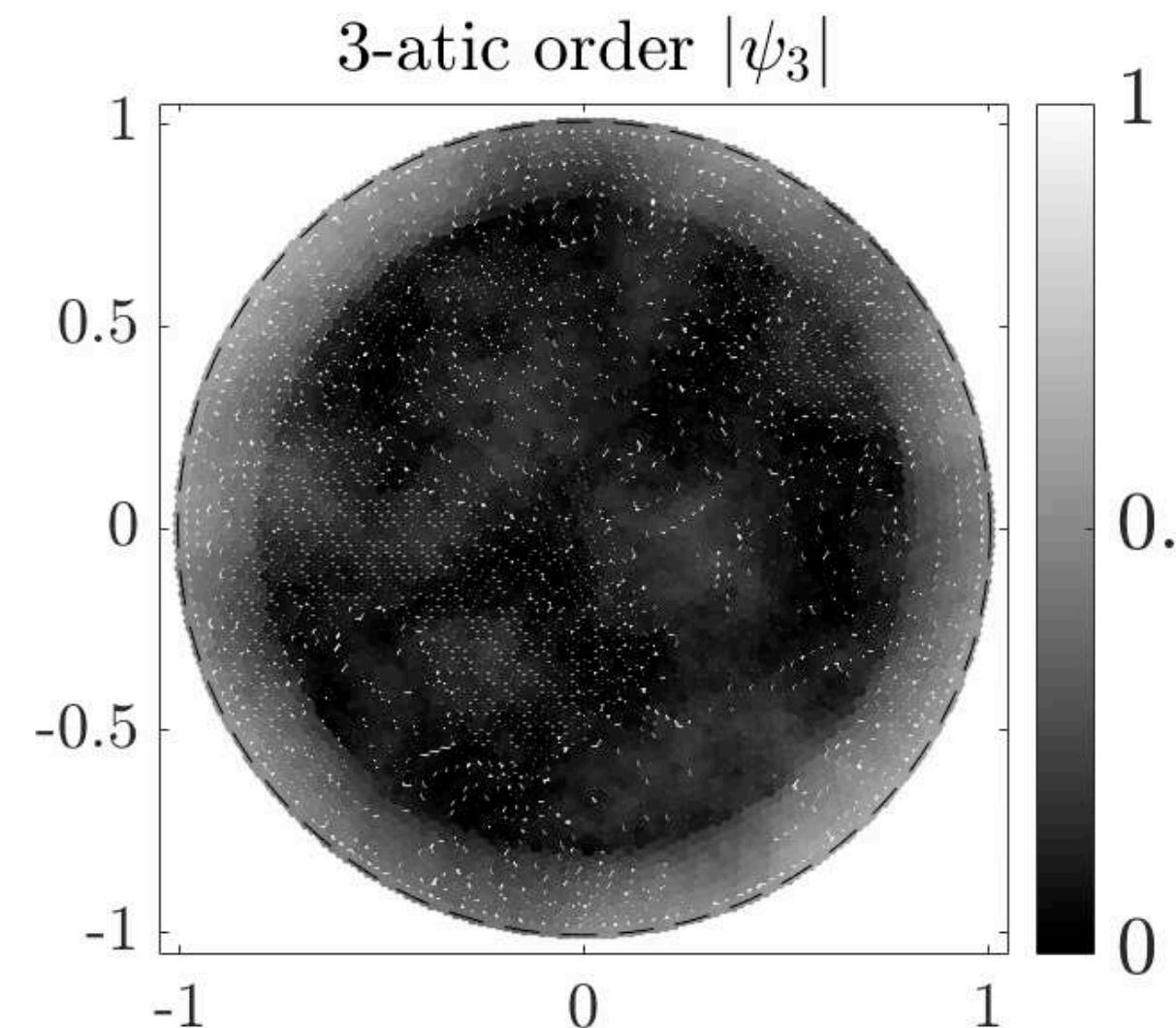
Landau-Brazovskii-Swift-Hohenberg



Relaxation dynamics



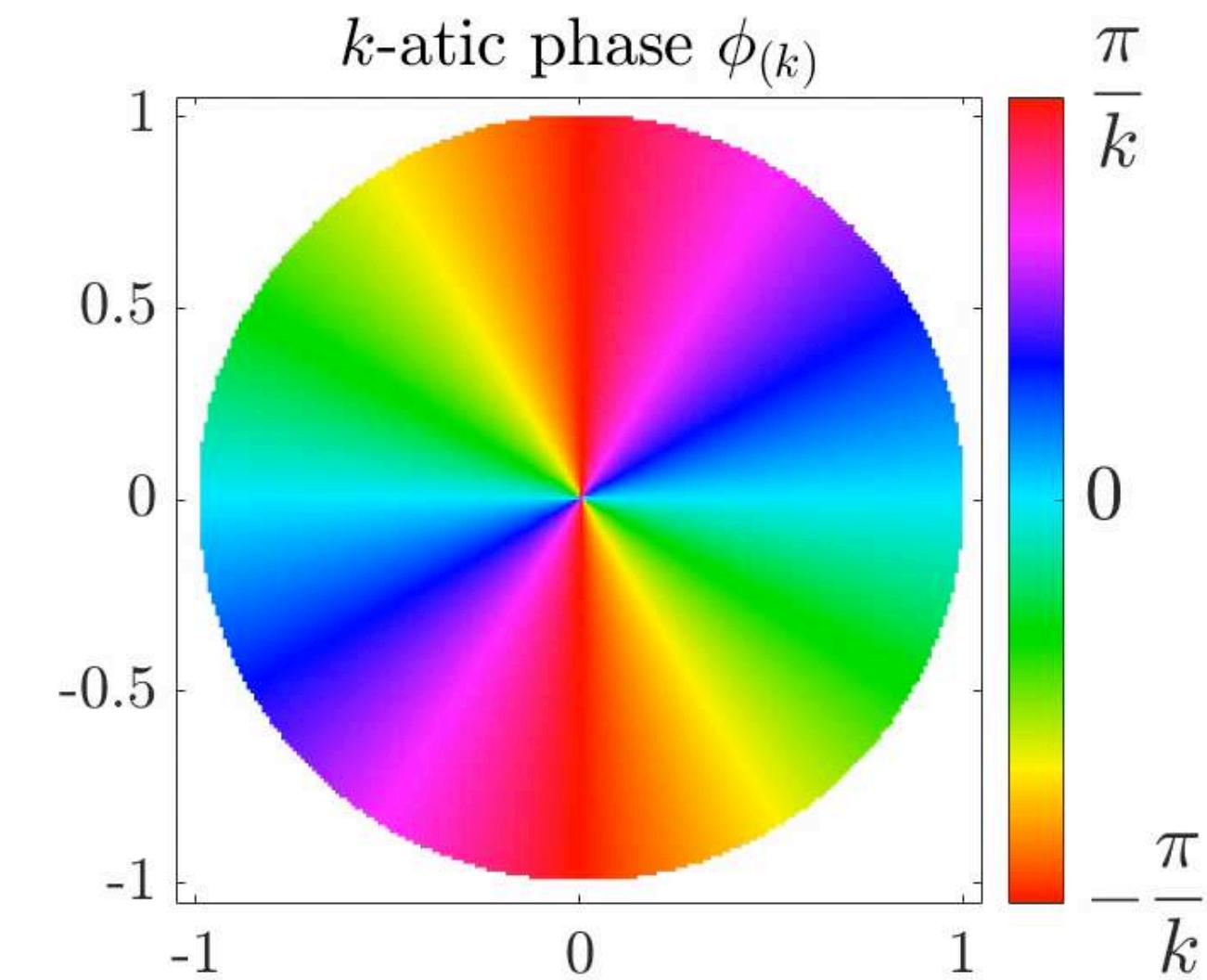
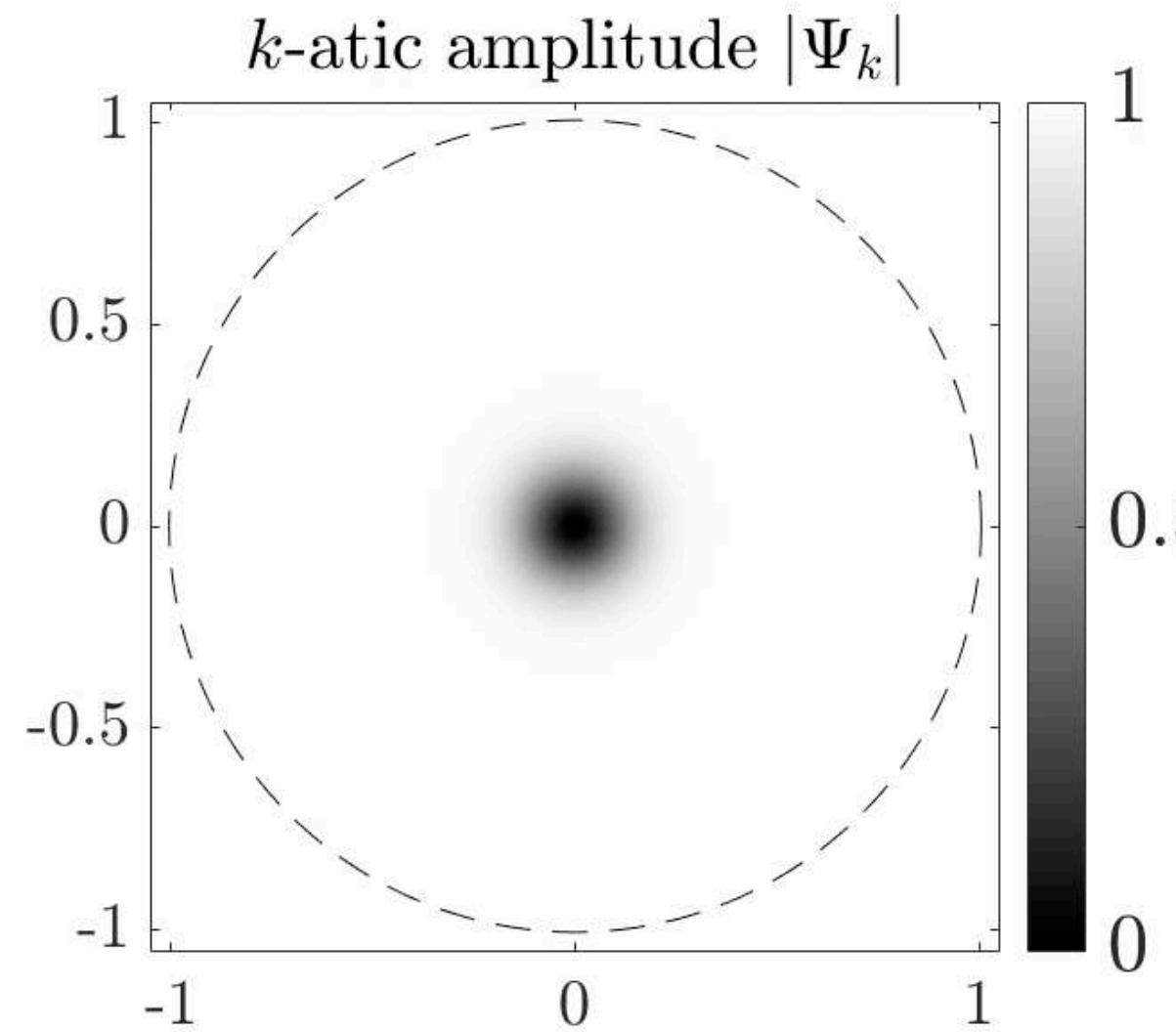
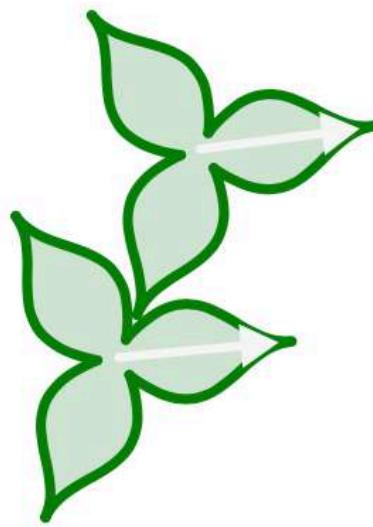
**Continuum
theory**



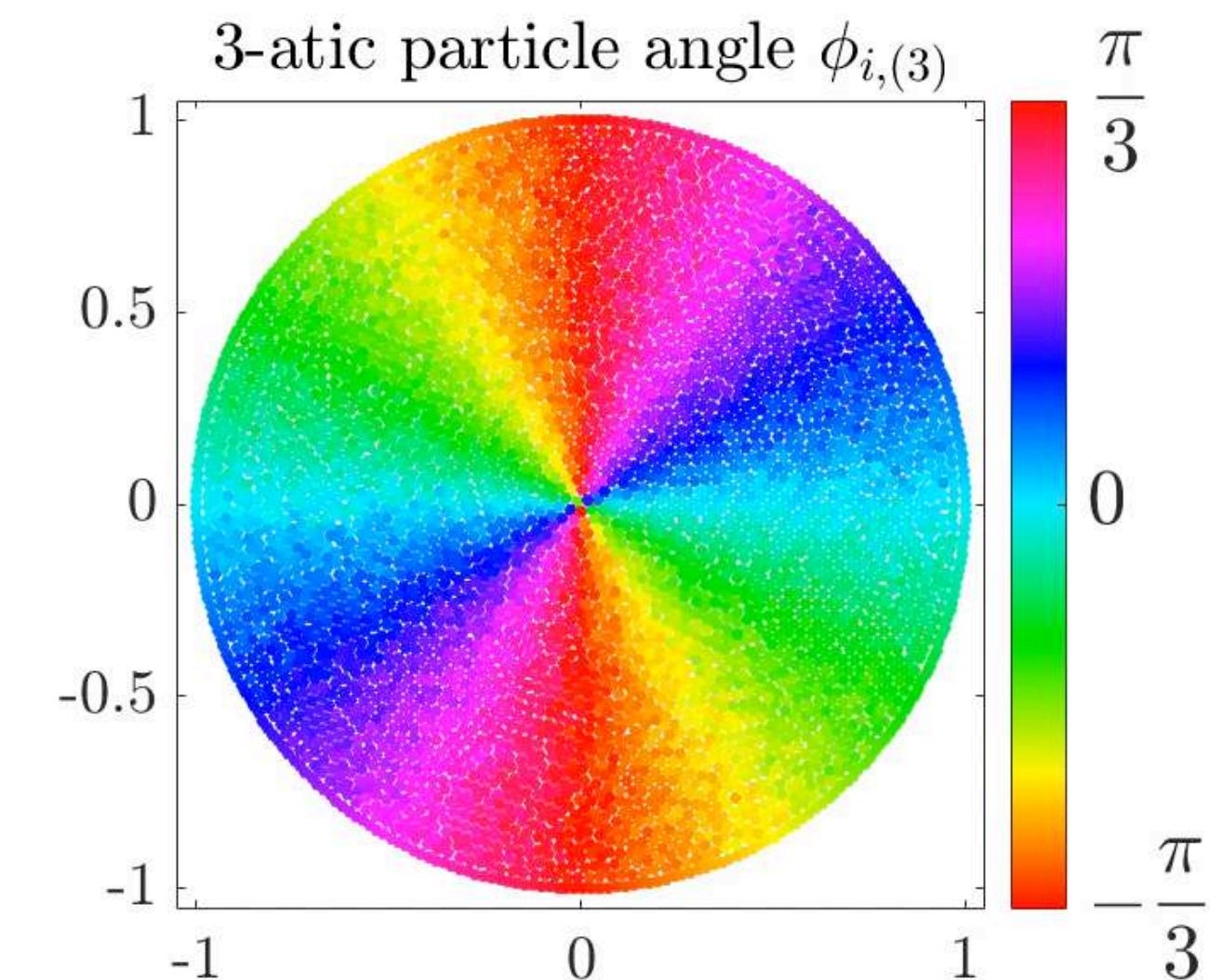
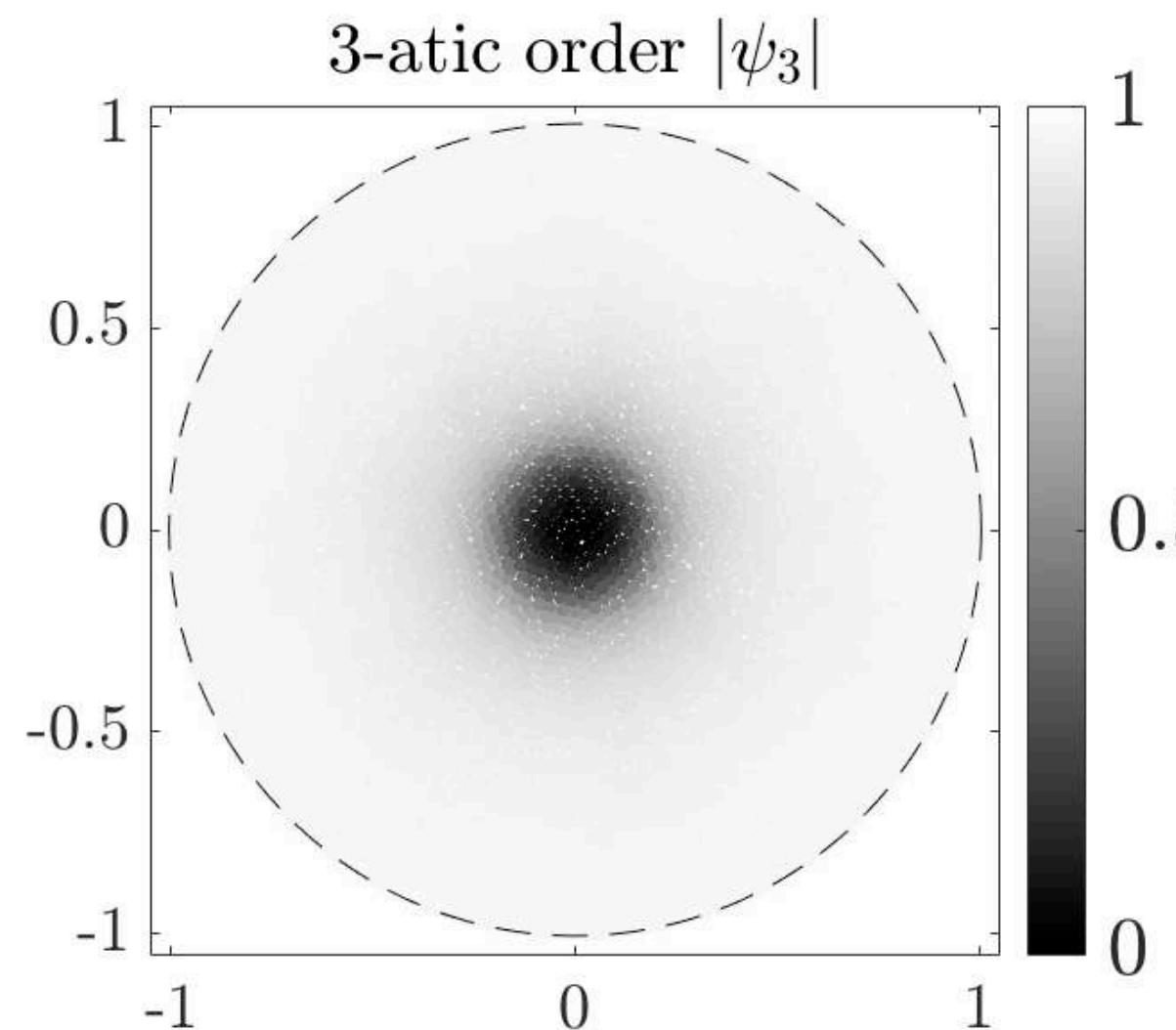
**Particle
simulation**



Defect splitting



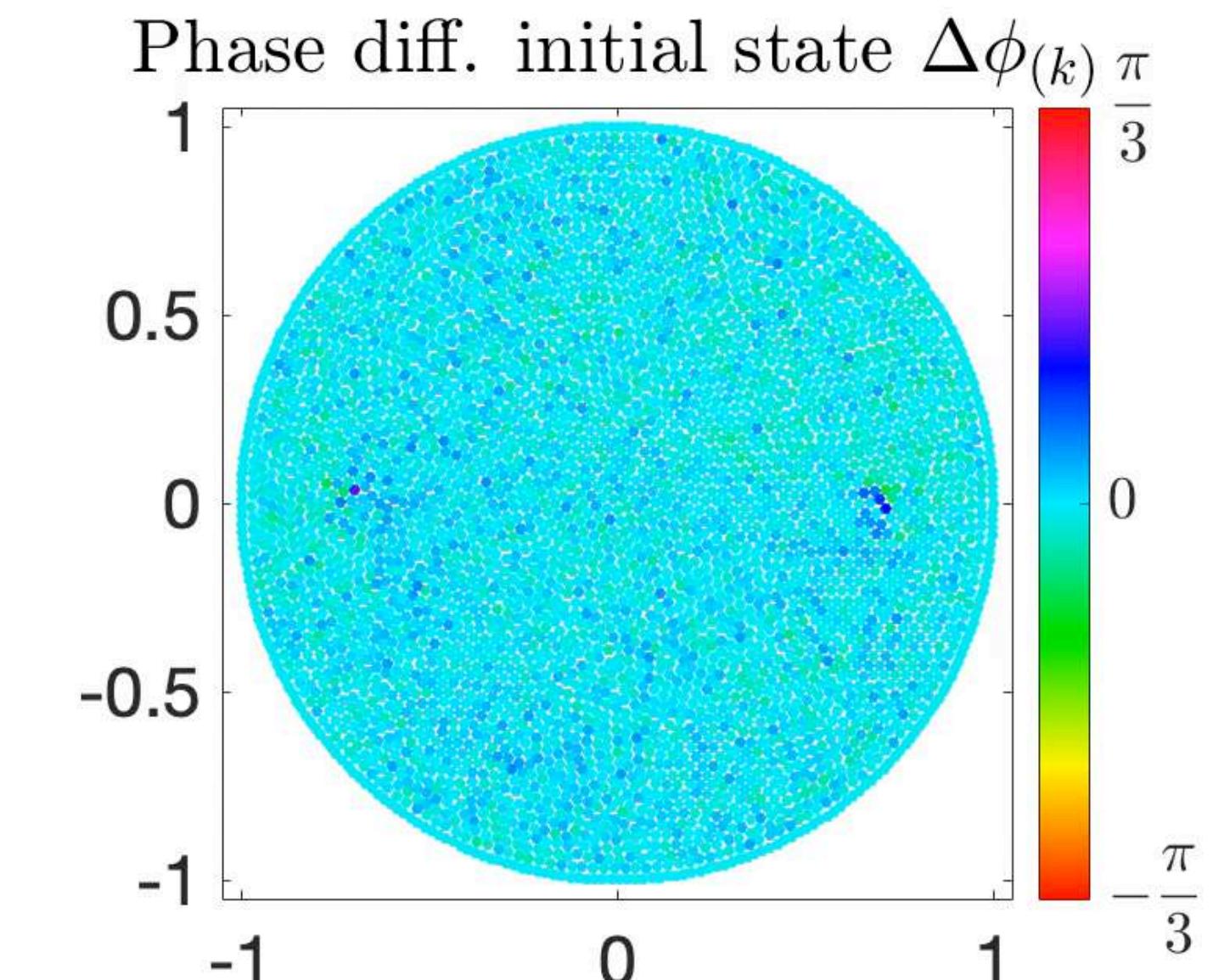
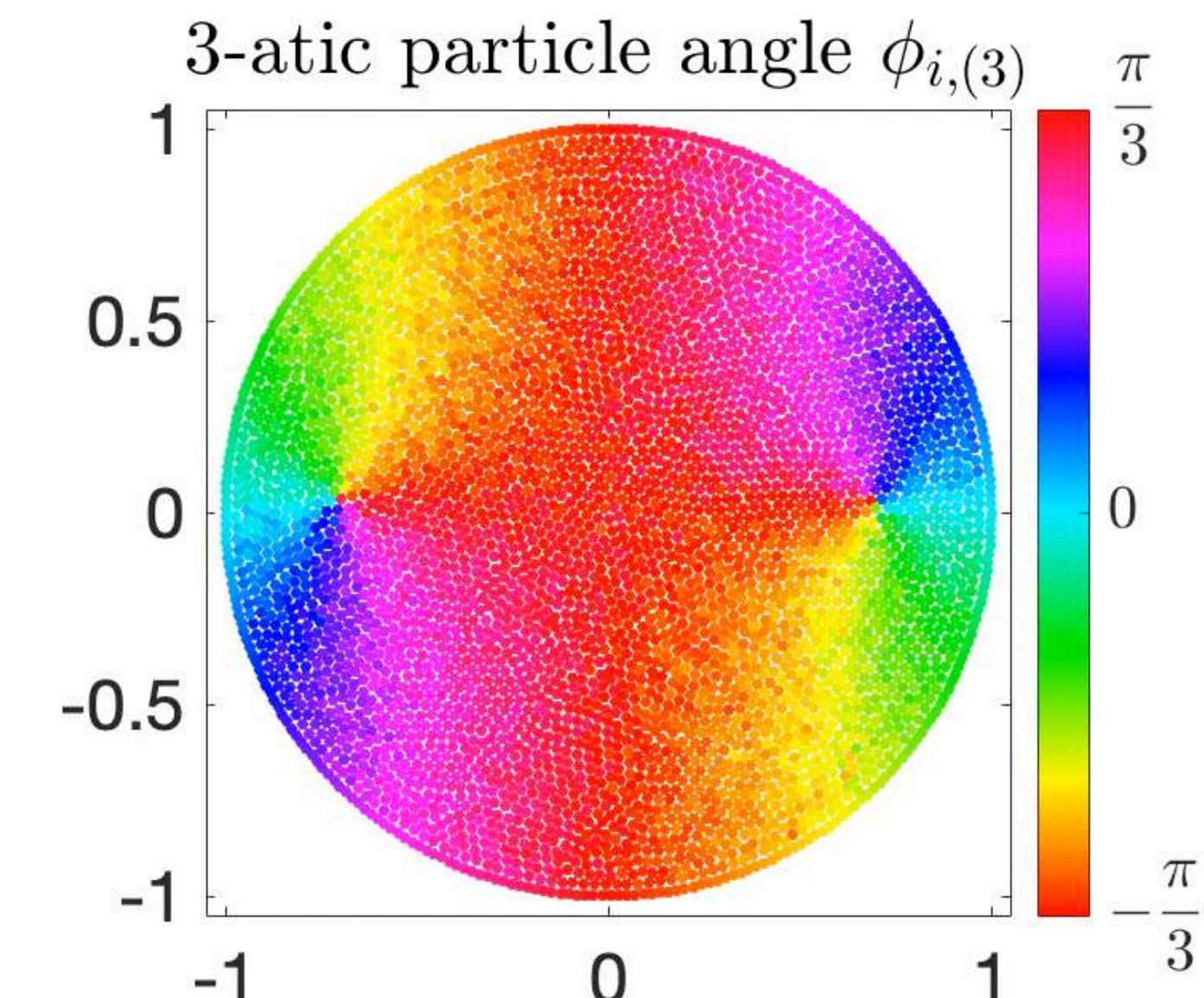
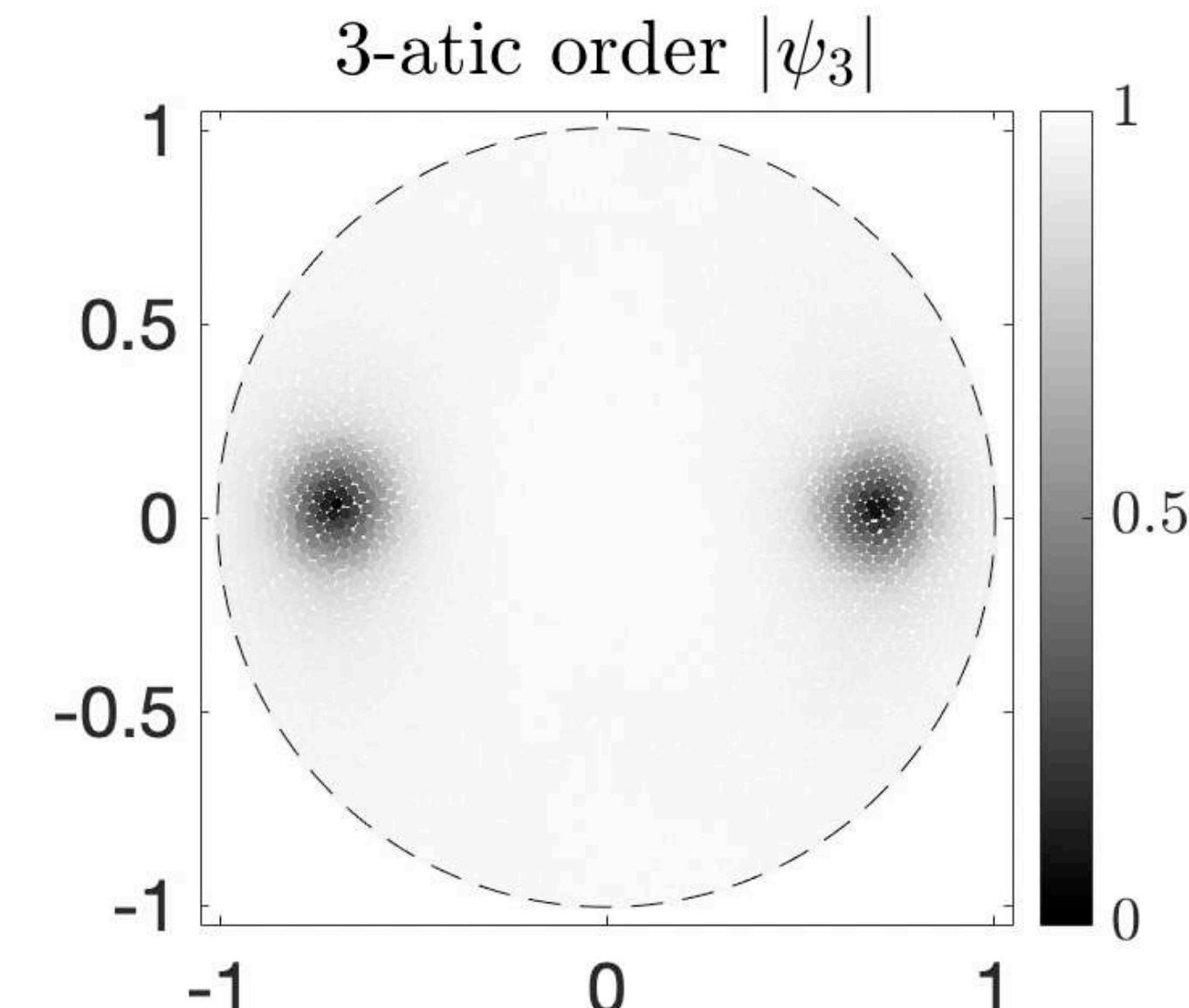
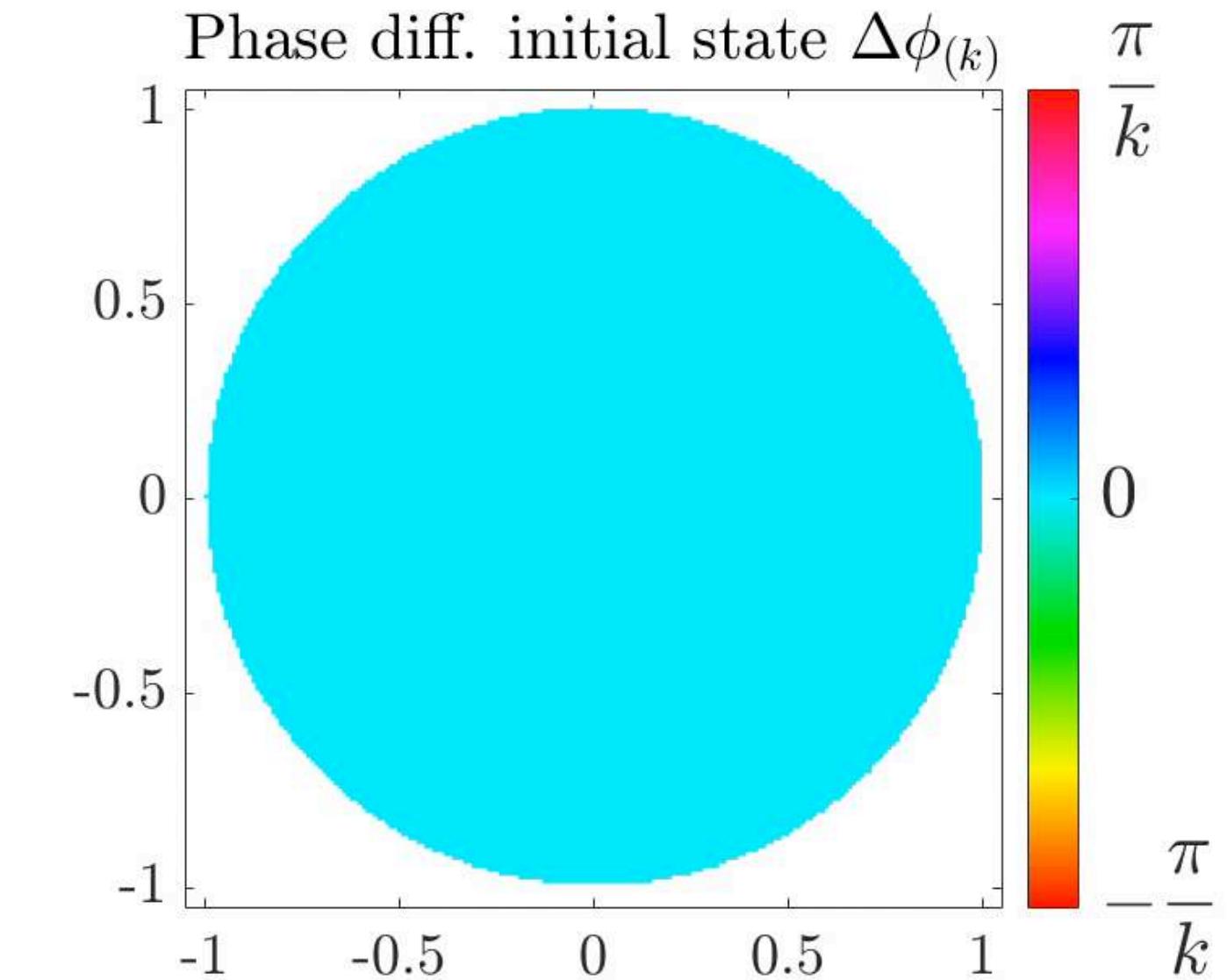
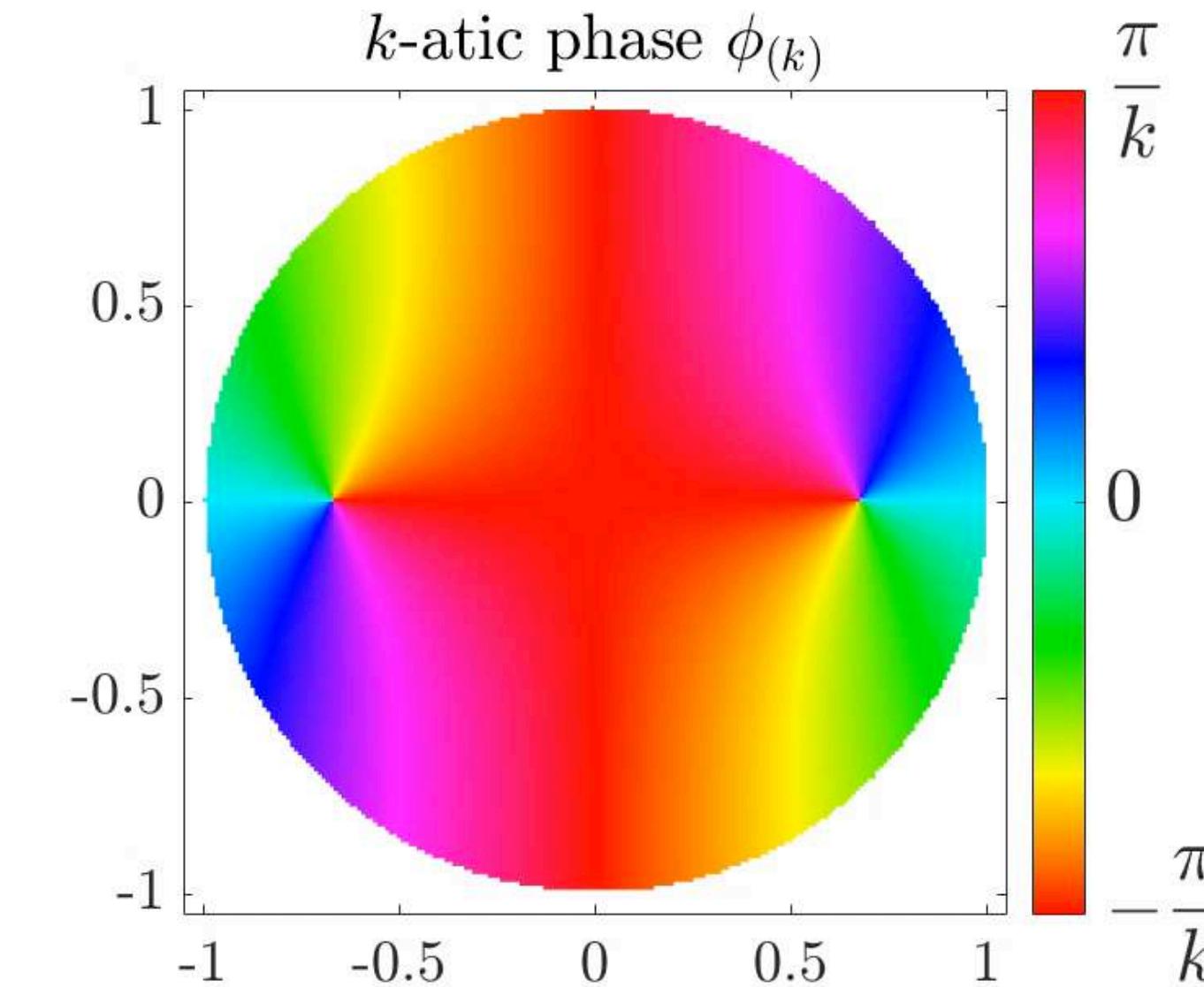
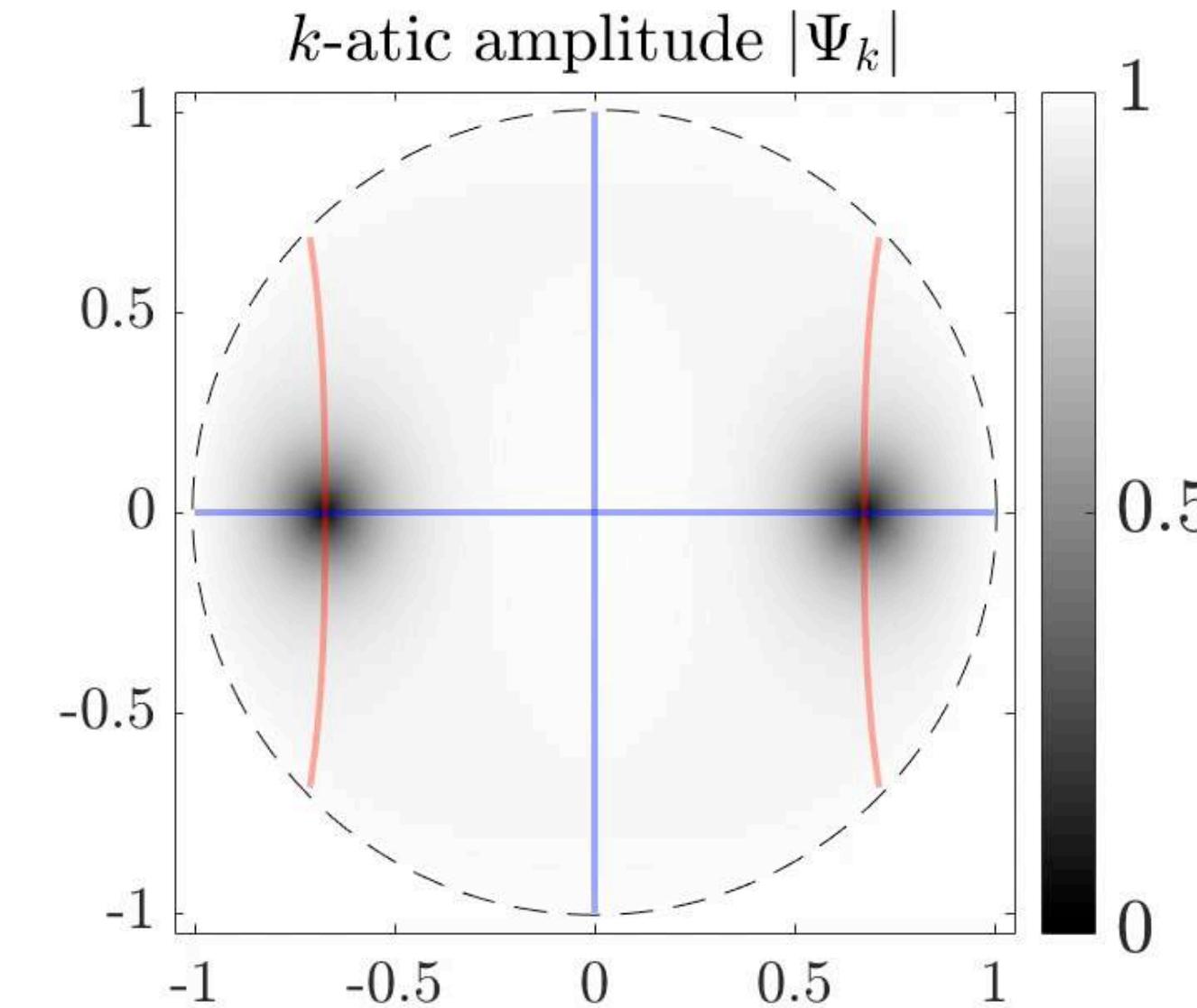
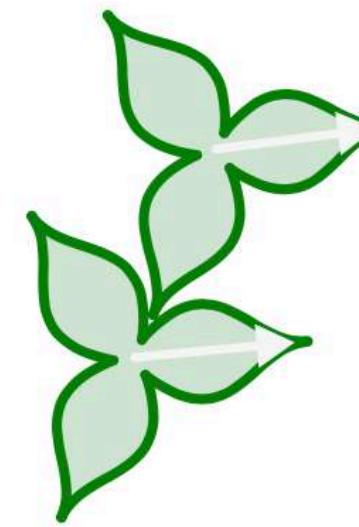
**Continuum
theory**



**Particle
simulation**

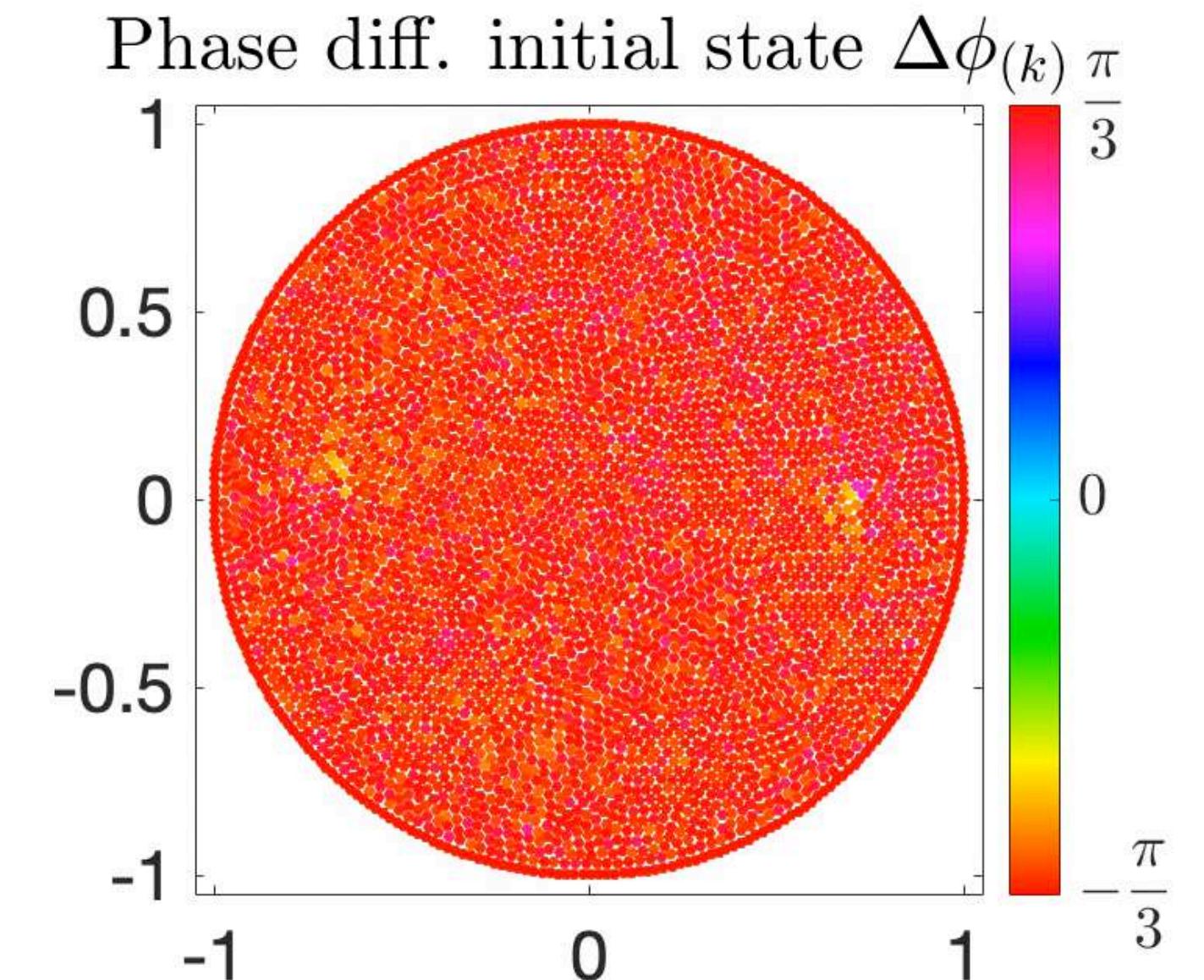
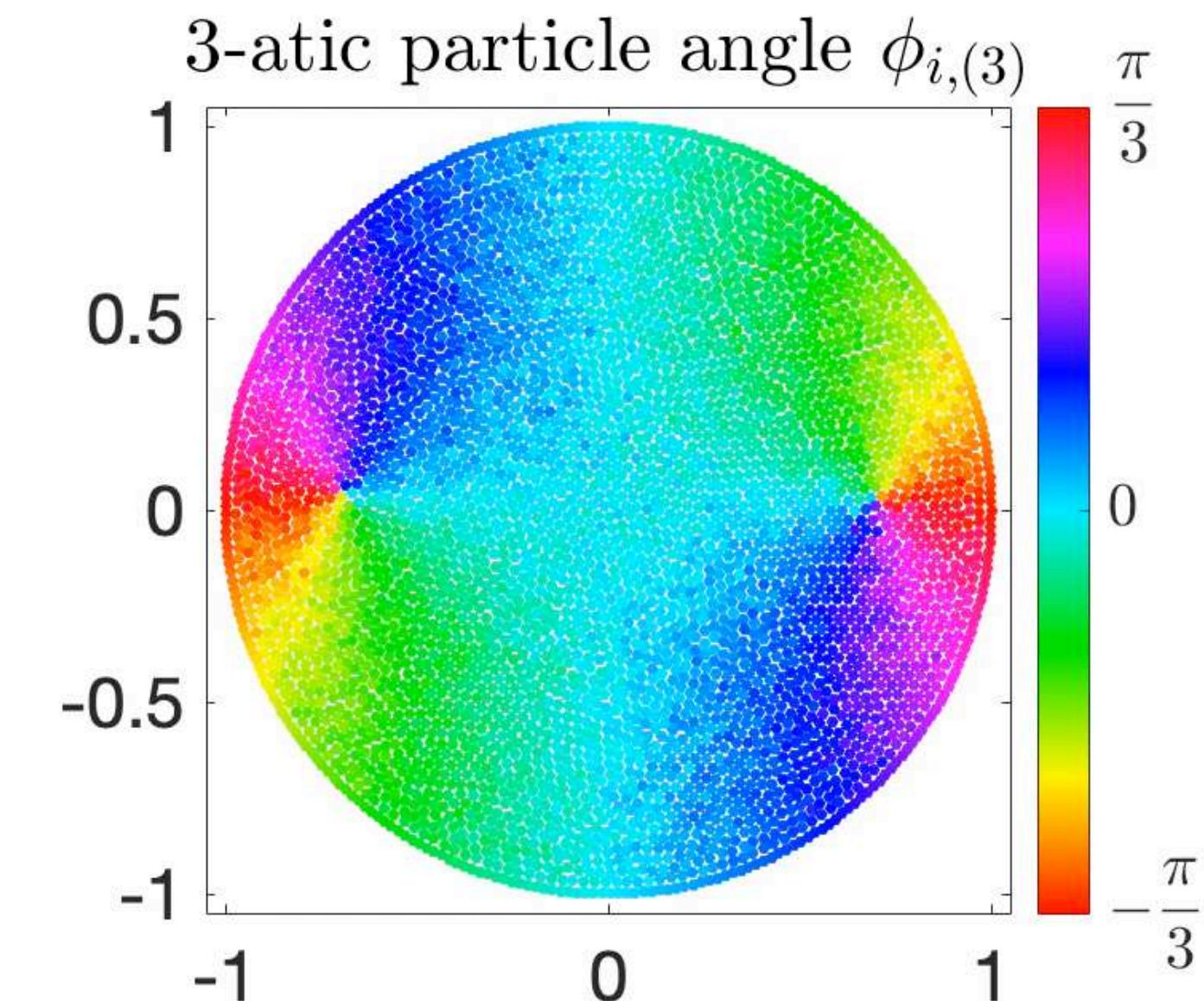
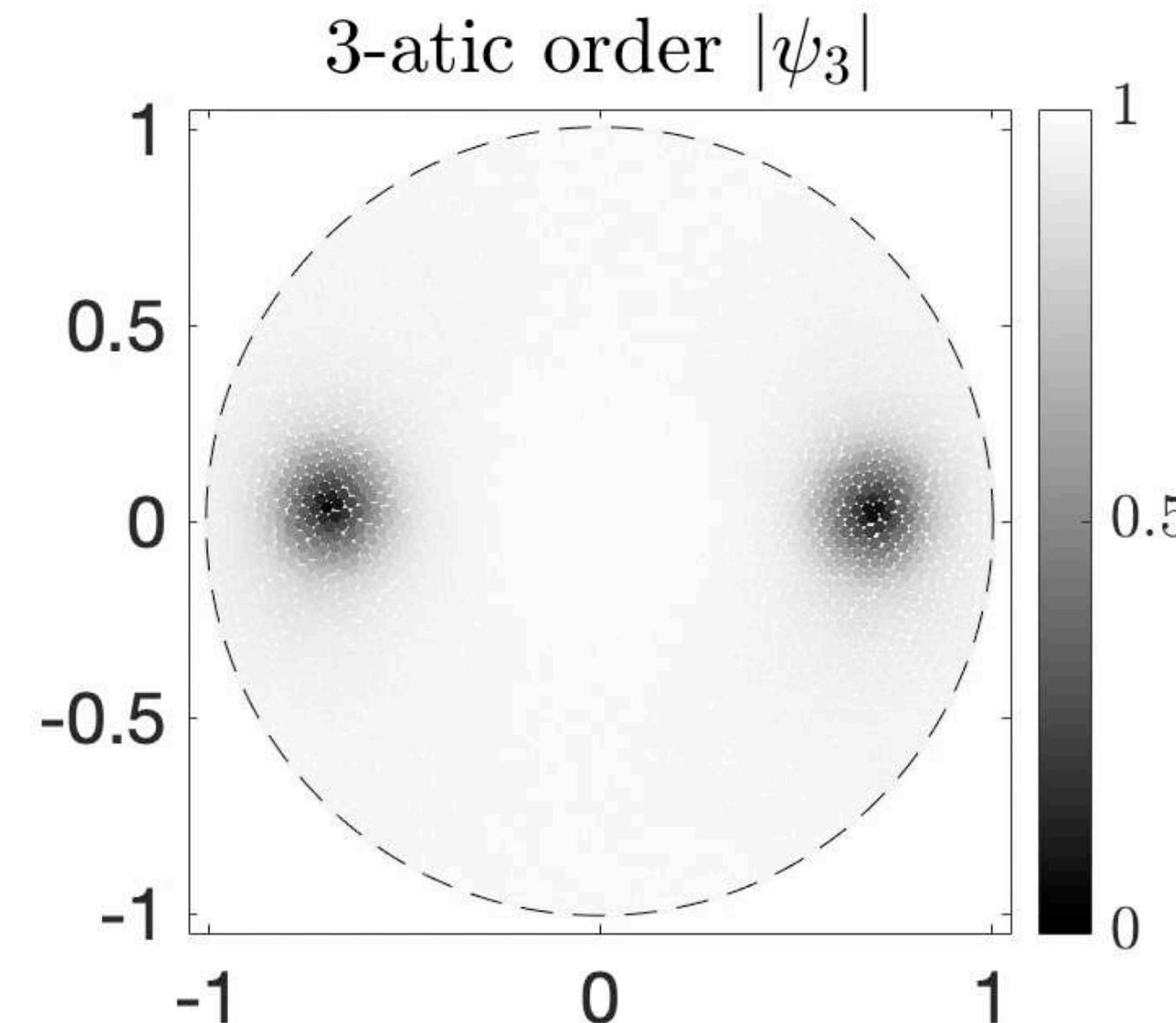
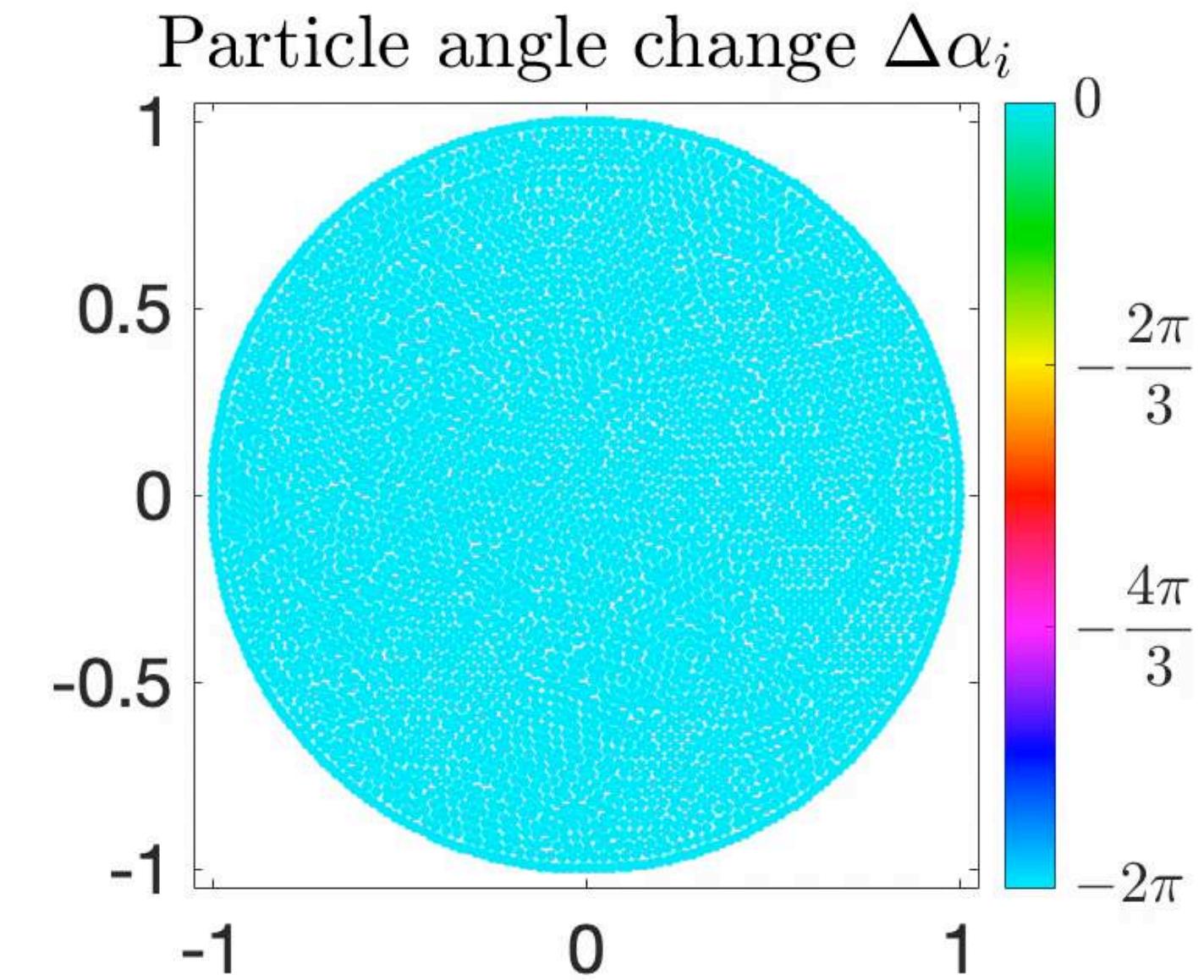
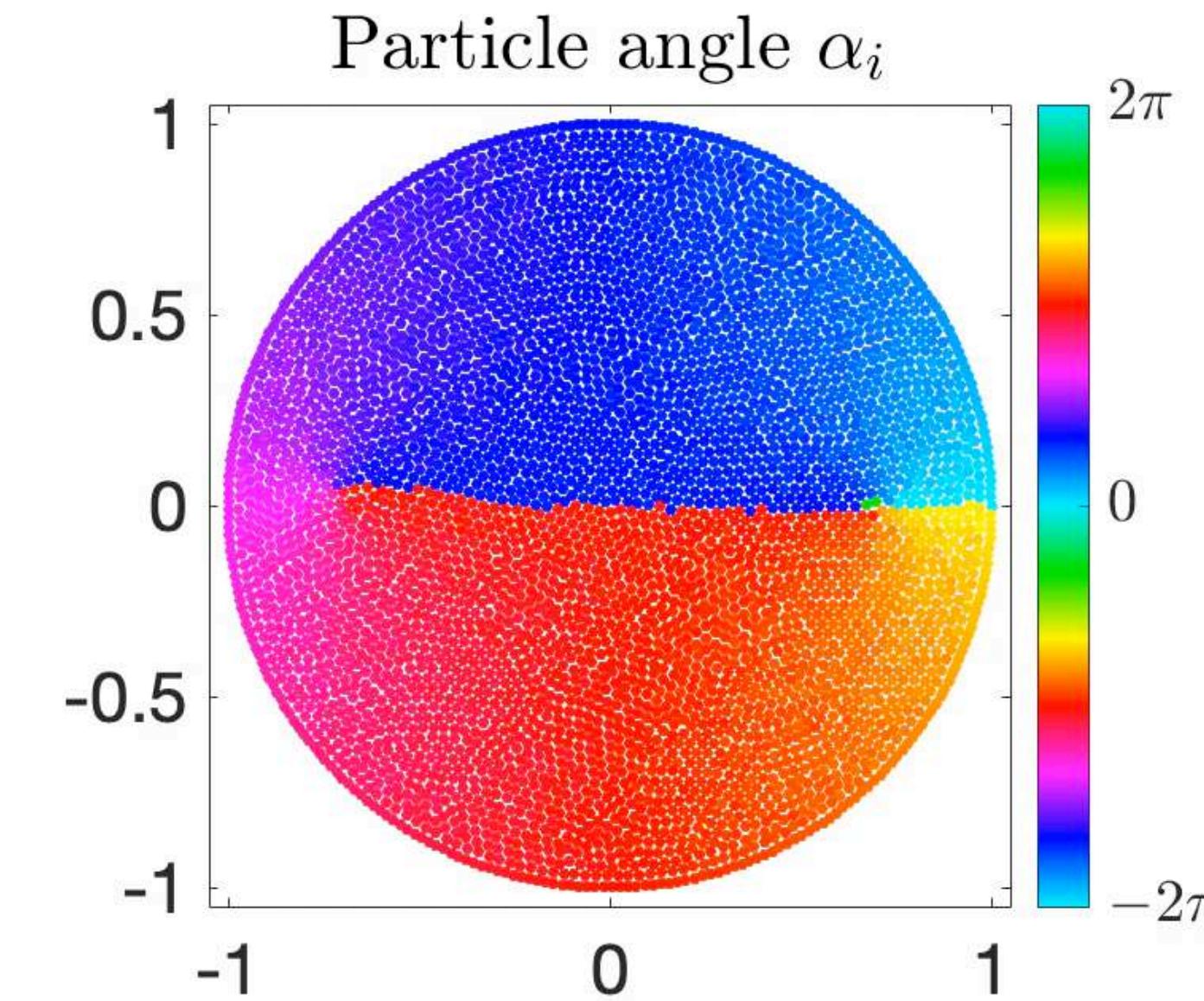
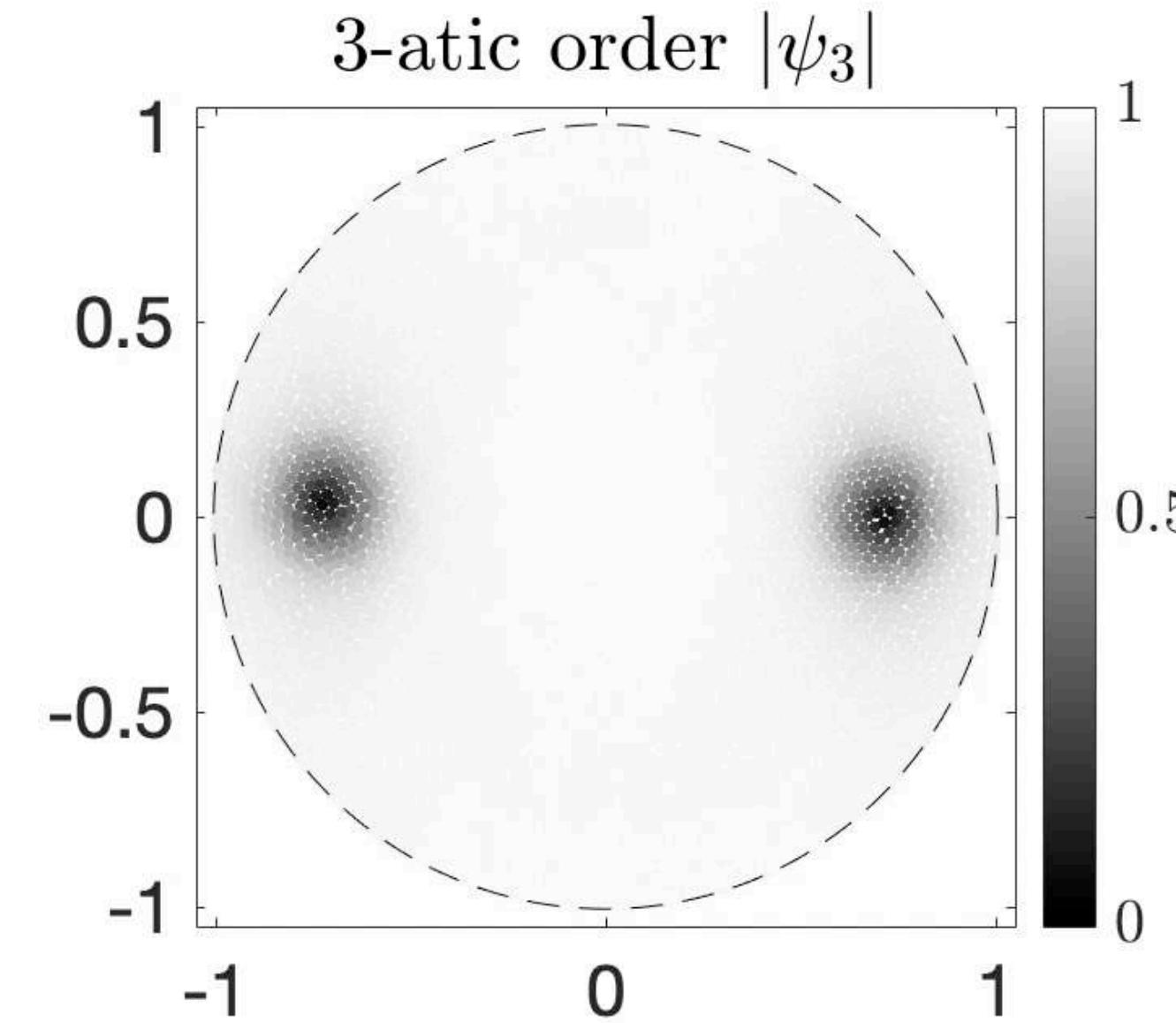
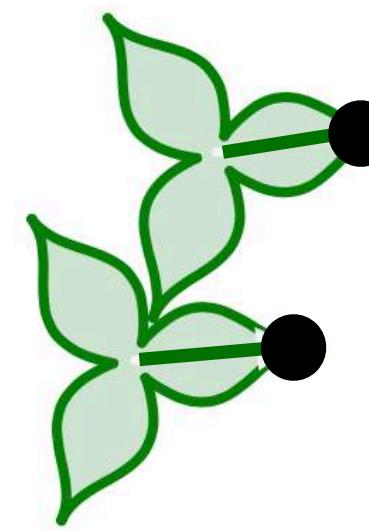
Defect braiding

[Time-dependent boundary](#)



Defect braiding

'anyon-like'





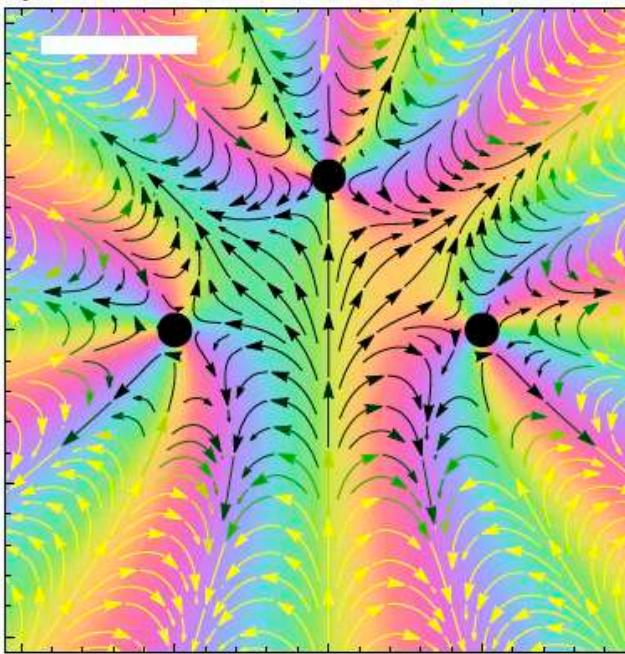
More analogies

PHYSICAL REVIEW RESEARCH 6, 043039 (2024)

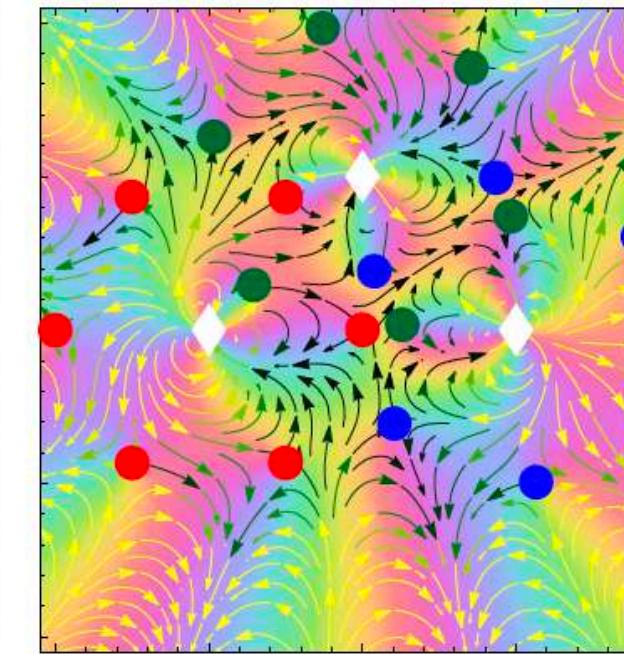
PHYSICAL REVIEW FLUIDS 6, 064702 (2021)

Harmonic flow-field representations of quantum bits and gates

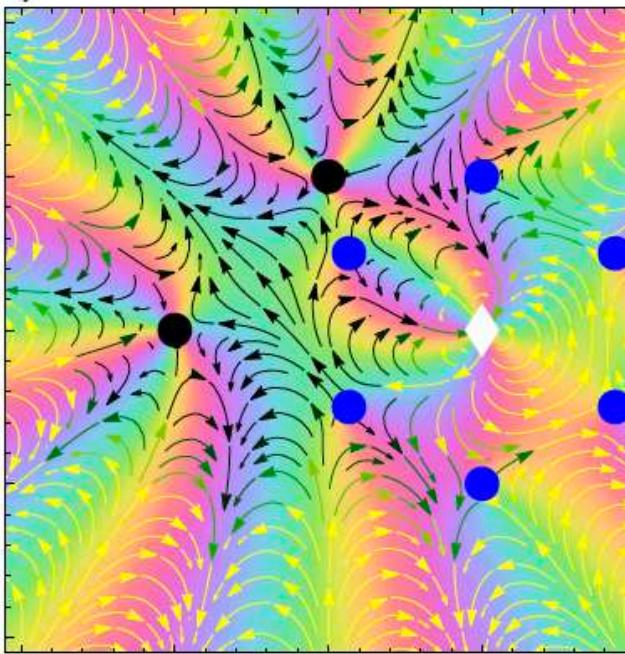
Vishal P. Patil,^{1,2,3,*} Žiga Kos,^{1,3,4,5} and Jörn Dunkel^{1,3,†}



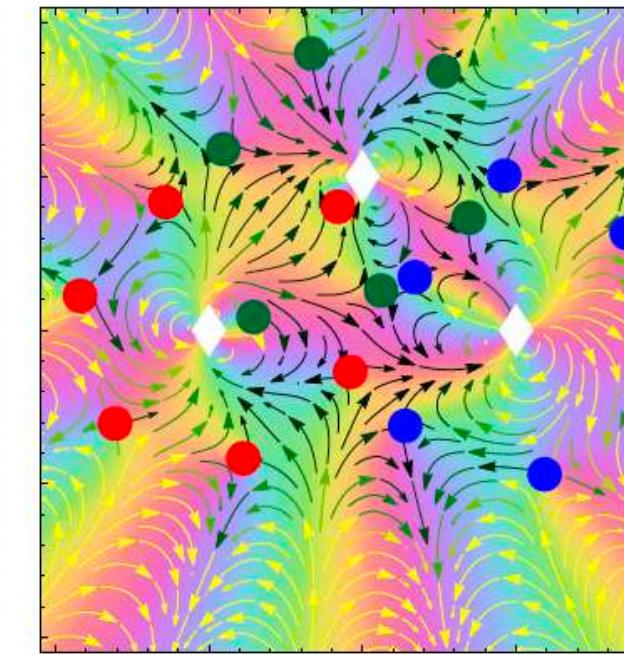
$|111\rangle$



$QFT|111\rangle$



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

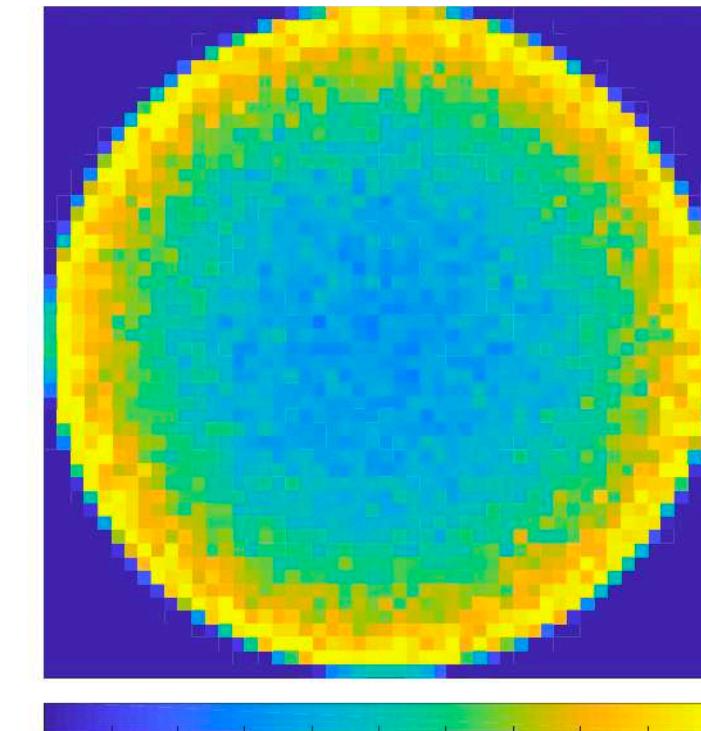
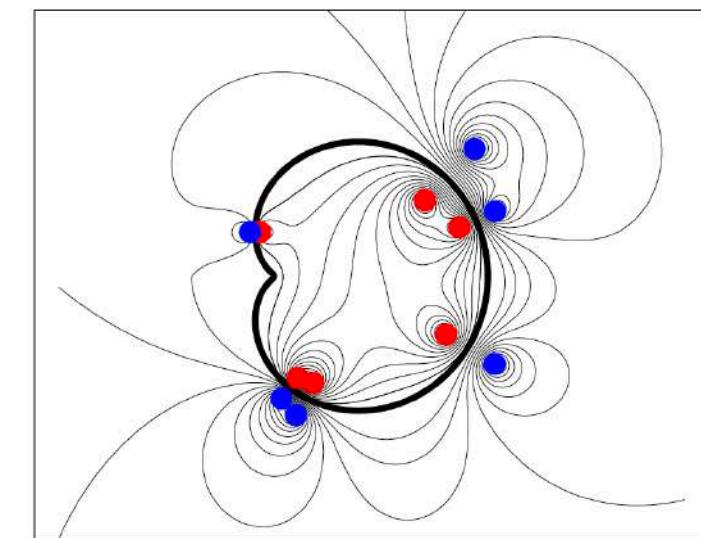
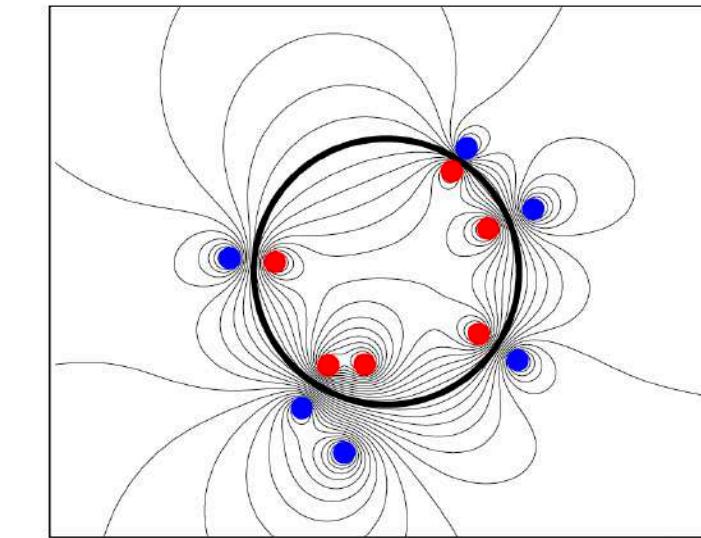


$QFT\left(\frac{1}{\sqrt{2}}|11\rangle \otimes (|0\rangle + |1\rangle)\right)$

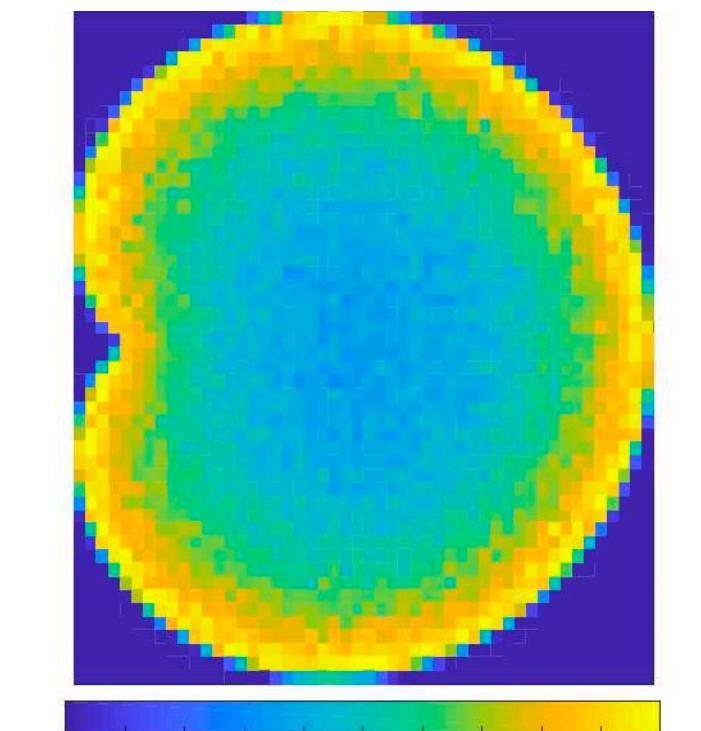
Mapping Hilbert space vectors to 2D compressible flow patterns

Chiral edge modes in Helmholtz-Onsager vortex systems

Vishal P. Patil and Jörn Dunkel^{1,3}



0 Probability density 0.5



0 Probability density 0.5

Real space argument: edge modes from boundary conditions