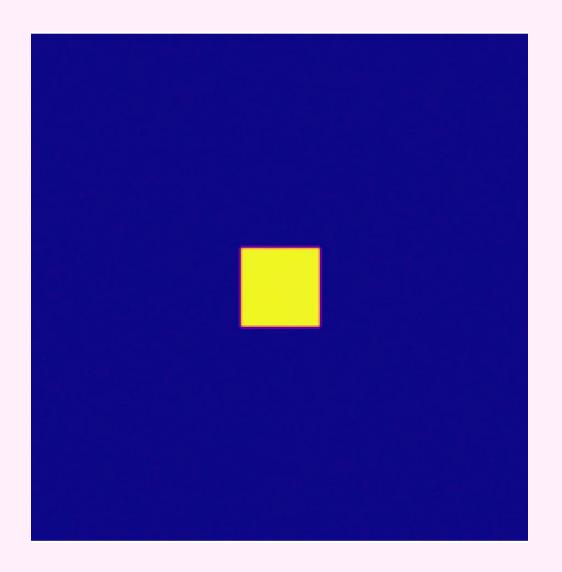
Turing patterns in reaction-diffusion systems

By Carolina and Sophia



Overview

Theory

Research question & motivation

Results: Pattern production

Results: Minkowski measures and functionals

Validity of results: Pearson comparison

Performance evaluation

Theory: Turing Patterns



→ all due to reaction and diffusion interplay!

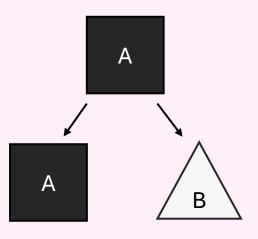


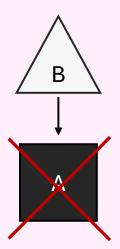














Images from Pearson's Complex patterns in a Simple System (1993)

Theory: Turing Patterns



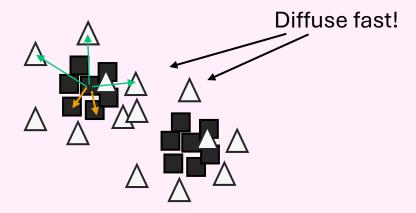
→ all due to reaction and diffusion interplay!







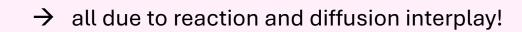


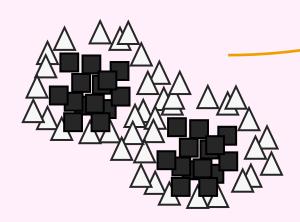




Images from Pearson's Complex patterns in a Simple System (1993)

Theory: Turing Patterns







Images from Pearson's Complex patterns in a Simple System (1993)

Theory: Grey-Scott model











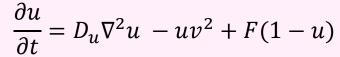






u - activator (A)

v - inhibitor (B)



$$\frac{\partial v}{\partial t} = D_v \nabla^2 v + u v^2 - (F + k)v$$

F - feed rate

k - kill rate

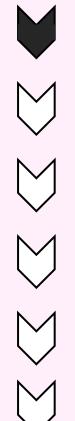


Solve using 5-point stencil in space dimension & forward Euler in time dimension

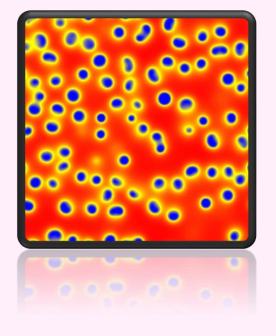




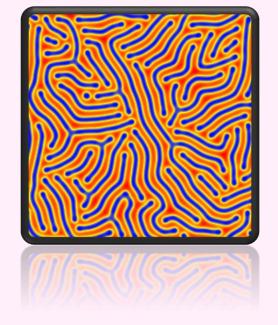




How can we quantify patterns?



VS



Theory: Minkowski Measures

Apply measures at each threshold grey level (1-256)





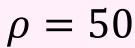














 $\rho = 100$



$$\rho = 150$$



$$\rho = 200$$



White pixel area



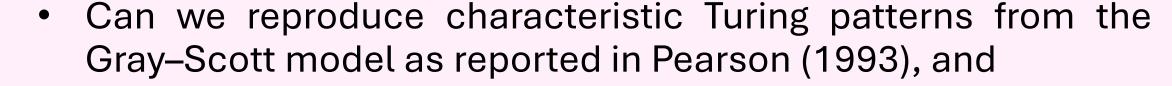


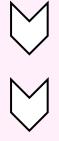
Boundary length Euler characteristic

Research question:









 Do our quantitative characterizations by Minkowski functionals correspond to experimental values established by Mecke (1996)?

Motivation

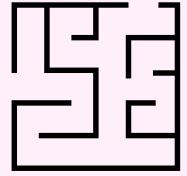




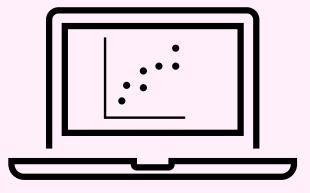








Turing patterns can explain pattern formation in nature!

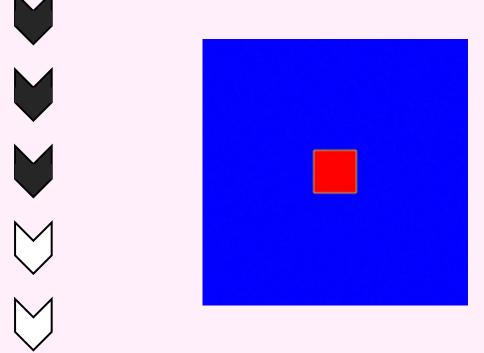


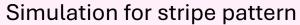
Little research on reproducing Mecke on modern simulations

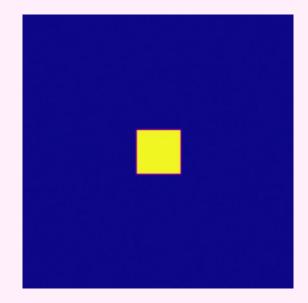
Results

The simulated system

Time-evolved simulations







Simulation for dotted pattern

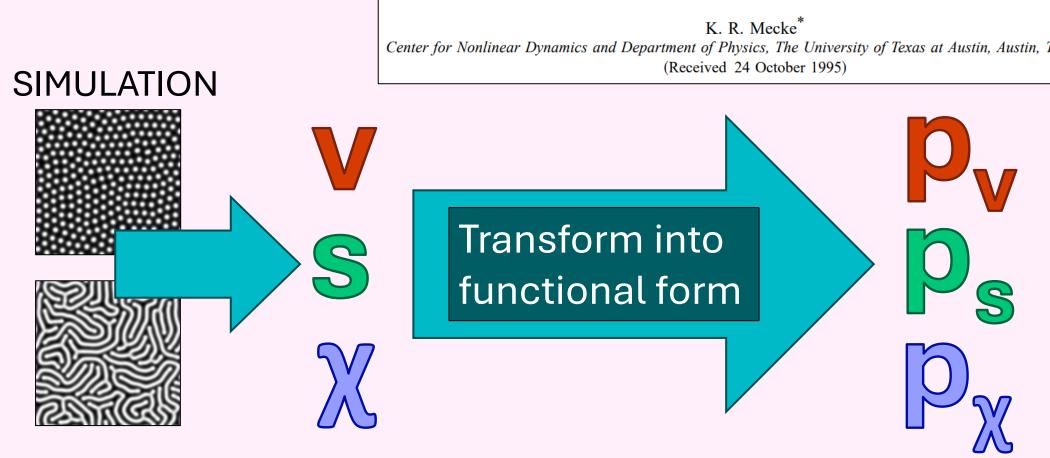
Results

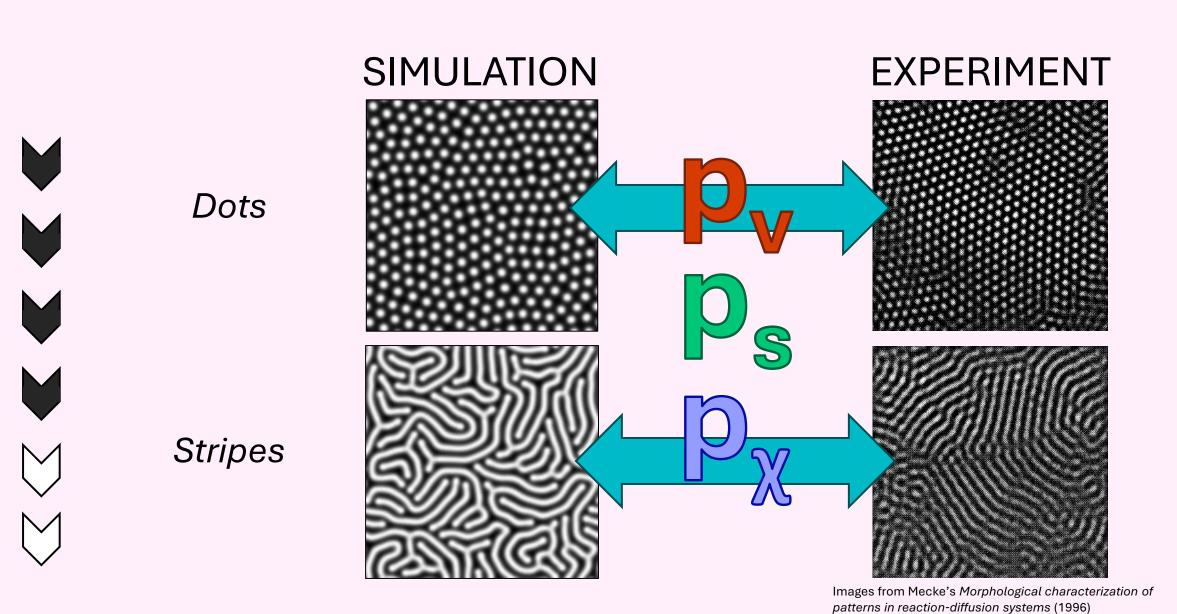
The Minkowski measures and functionals

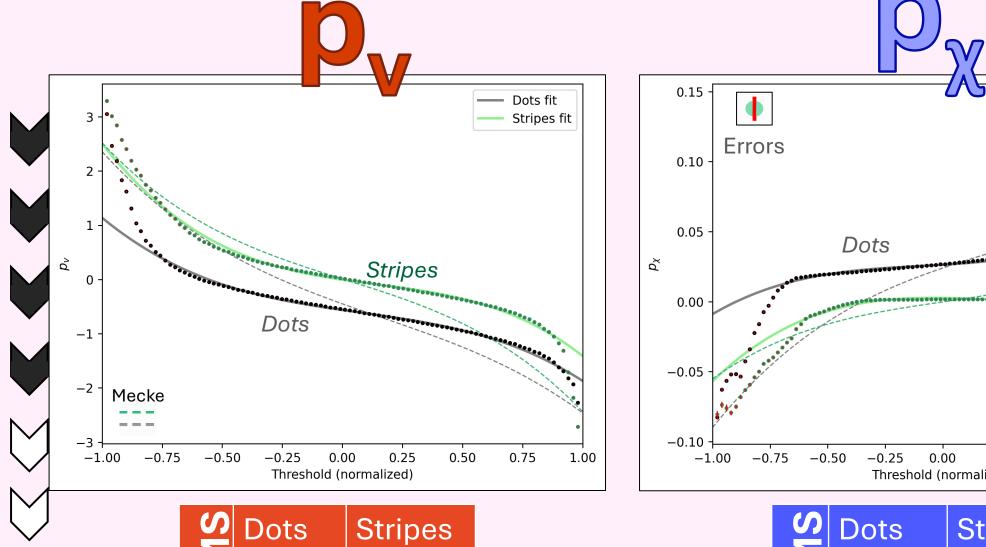
Comparing to literature

Morphological characterization of patterns in reaction-diffusion systems

Center for Nonlinear Dynamics and Department of Physics, The University of Texas at Austin, Austin, Texas 78712 (Received 24 October 1995)

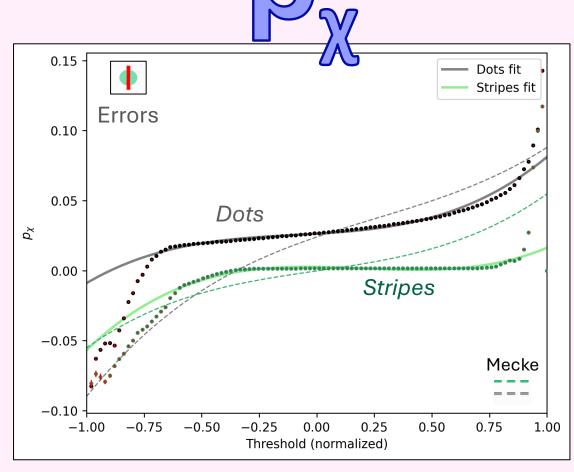




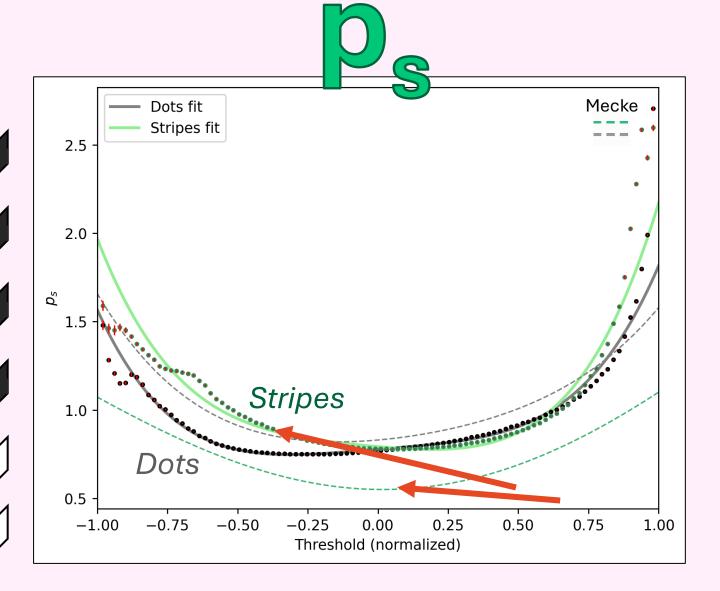


0.084

0.118



15	Dots	Stripes
A	0.176	0.137



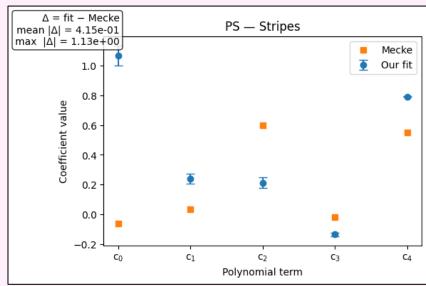
RMS vs Mecke

Dots

Stripes

0.131

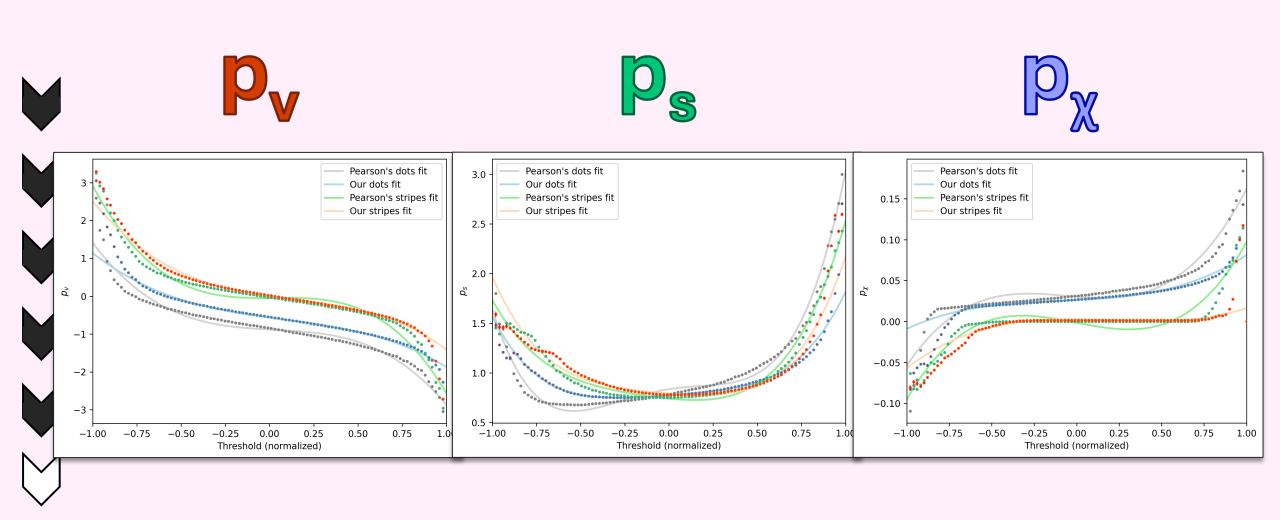
0.720



*Coefficient errors for all fits in appendix

Validity of simulation results

Comparing our functionals to Pearson's (1993)

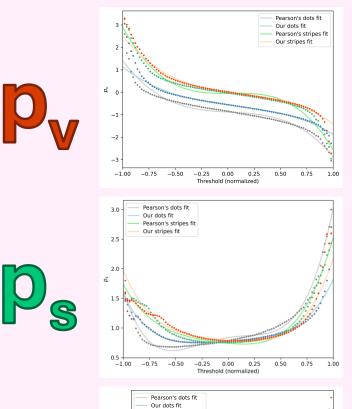


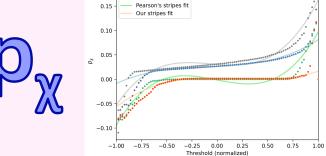
Statistical analysis:

Pearson's vs. our functionals

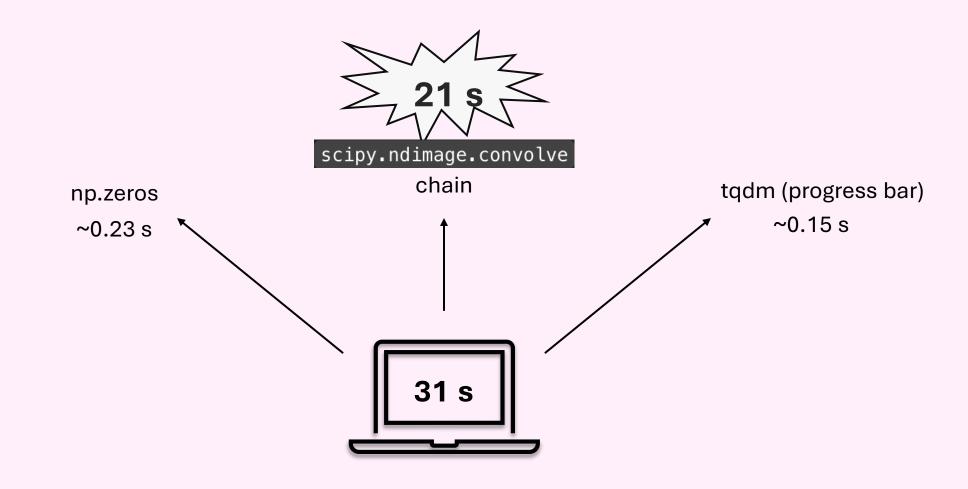
RMS		
Dots	Stripes	
0.131	0.080	
0.283	0.083	
0.258	0.290	

Orientation agreement		
Same-sign	Pearson corr. coeff.	
1.000	0.851	
0.698	0.531	
1.000	0.893	

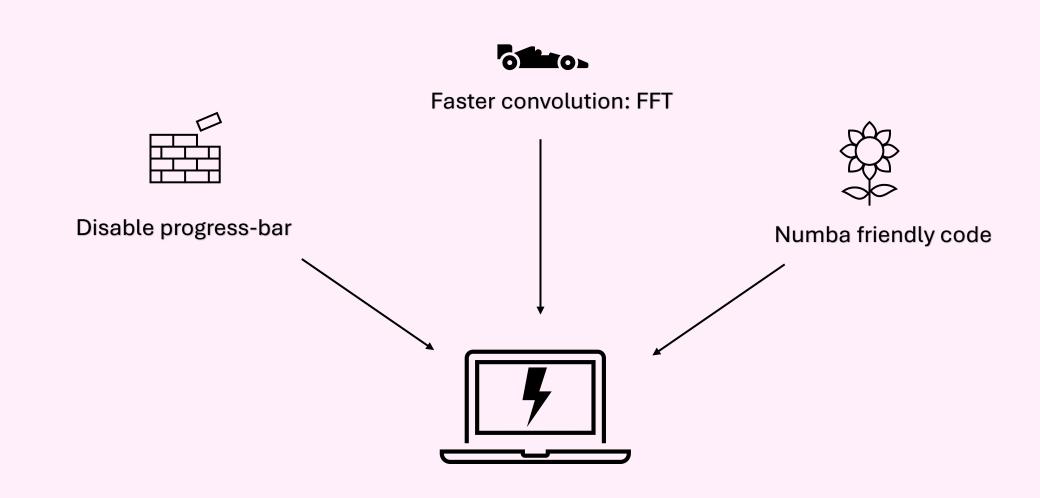




Performance evaluation

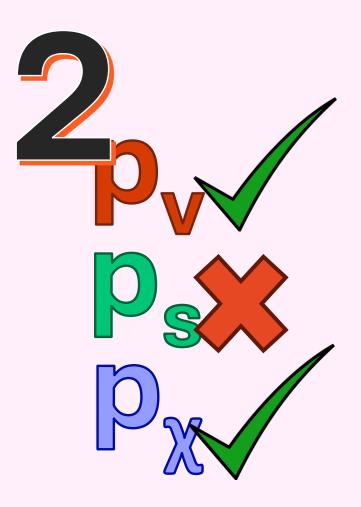


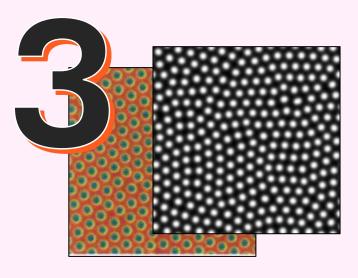
Performance evaluation

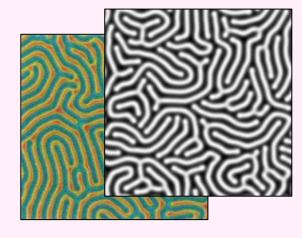


Conclusion



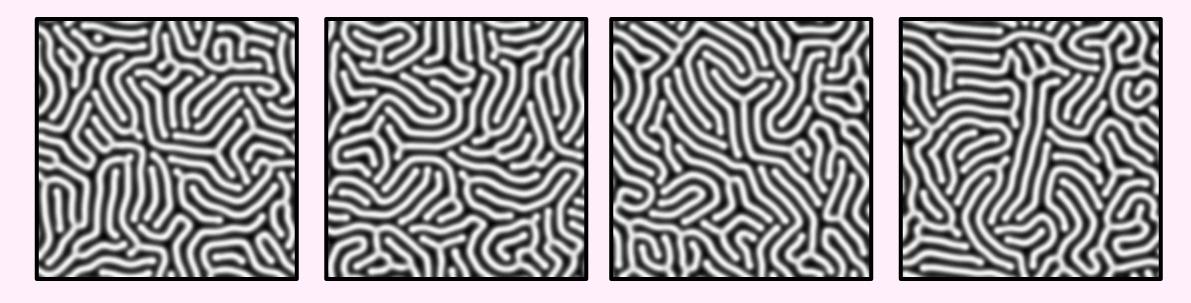






Thank you for listening!

Extra slides: Errors



Different RNG seeds for each run – 20 runs, mean and std error for p, v and chi, as well as p_v, p_s, p_chi

Polynomial analysis

$$NRMS_{range} = \frac{\sqrt{\frac{1}{N} \sum_{i=1}^{N} [P_1(r_i) - P_2(r_i)]^2}}{\max_{r} P_2(r) - \min_{r} P_2(r)}$$

Normalised RMS to compare p chi with the others: a dimensionless 'fraction-of-full-height' error

$$NRMS_{range} = \frac{\sqrt{\frac{1}{N} \sum_{i=1}^{N} \left[P_1(r_i) - P_2(r_i) \right]^2}}{\max_{r} P_2(r) - \min_{r} P_2(r)}} \rho = \frac{\sum_{i=1}^{N} \left(\Delta_{ours}(r_i) - \overline{\Delta}_{ours} \right) \left(\Delta_{Mecke}(r_i) - \overline{\Delta}_{Mecke} \right)}{\sqrt{\sum_{i=1}^{N} \left(\Delta_{ours}(r_i) - \overline{\Delta}_{ours} \right)^2} \sqrt{\sum_{i=1}^{N} \left(\Delta_{Mecke}(r_i) - \overline{\Delta}_{Mecke} \right)^2}},$$

Pearson coefficient – measures linear relationship between two variables; shows us global consistency of fluctuations around the mean (mean-centered, d/n take into account vertical offsets).

Cosine angle between two sampled difference vectors

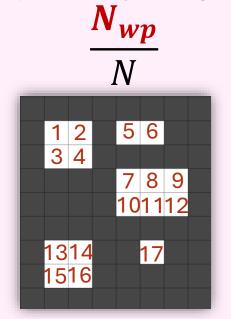
$$\Delta_{ours}(r) = P_{dots}(r) - P_{stripes}(r), \Delta_{Mecke}(r) = M_{dots}(r) - M_{stripes}(r)$$

$$f_{same} = \frac{1}{N} \sum_{i=1}^{N} \mathbf{1} [\Delta_{ours}(r_i) \Delta_{Mecke}(r_i) > 0],$$

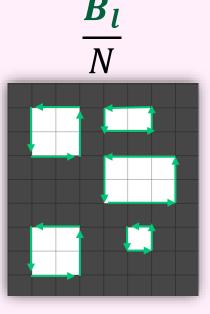
Fraction of thresholds whose orientation matches

Theory: Minkowski Measures

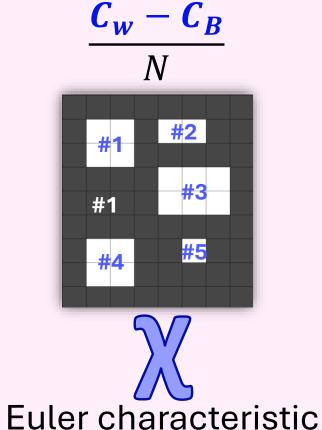
Geometric measures for morphological differences





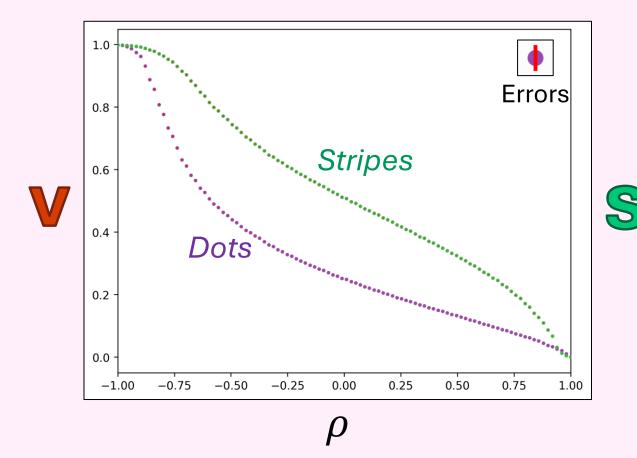


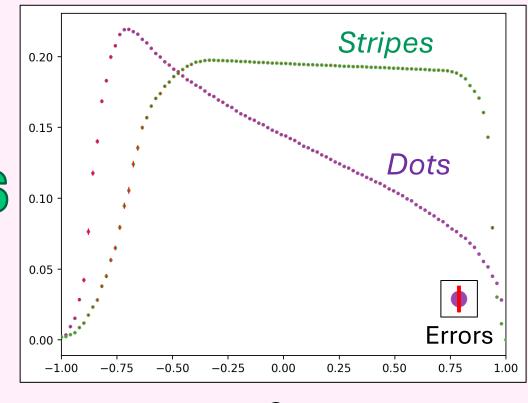




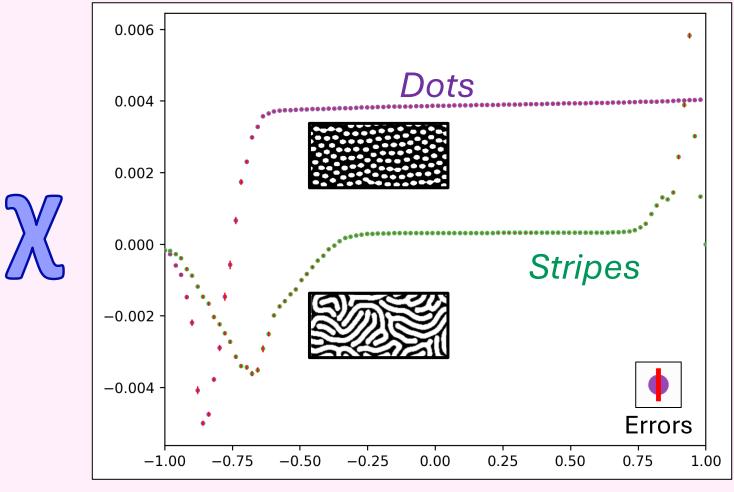
White pixel area

Boundary length





Difference in # of components



Extra slides: Functional forms

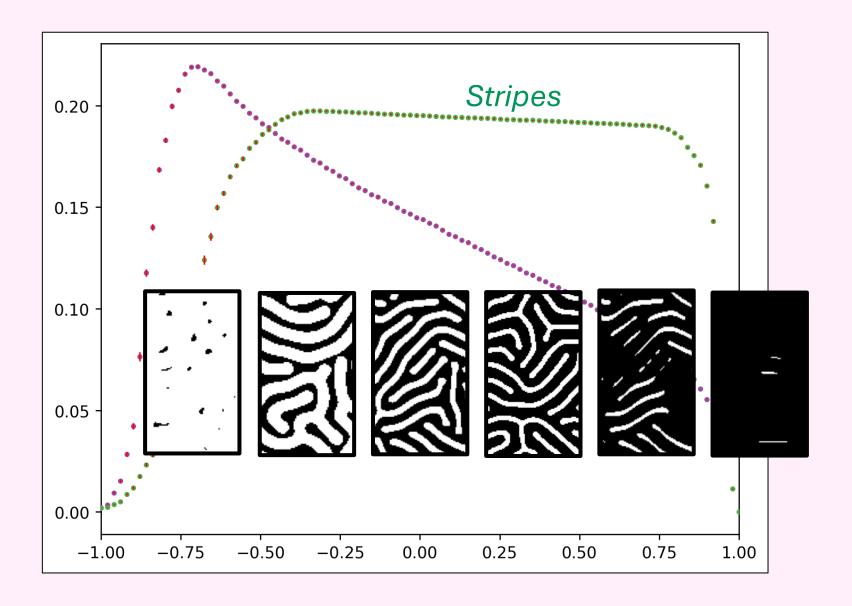
$$p_v(\rho) = \tanh^{-1}(2v - 1),$$

$$p_s(\rho) = \frac{s}{v(1-v)} = 4s \cosh^2[p_v(\rho)],$$

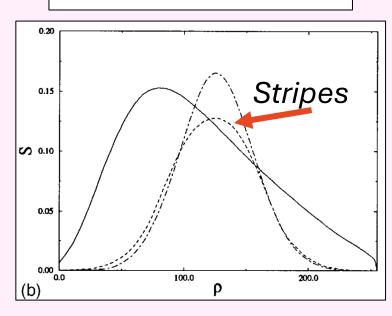
$$p_{\chi}(\rho) = \frac{\chi}{s},$$

Chosen by Mecke because the functionals occur in many fields of physics. Direct connection to reaction-diffusion systems unclear.

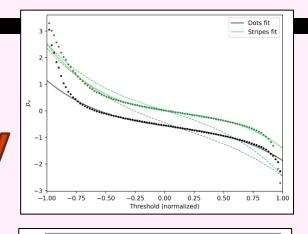
Allows for polynomial fitting – and coefficients that depend on experimental conditions

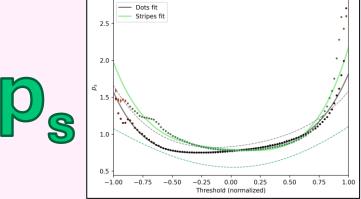


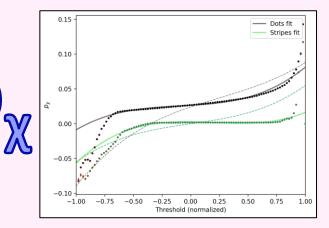
$$p_s(\rho) = \frac{s}{v(1-v)}$$



Shape due to inhomogeneities in experimental photos?



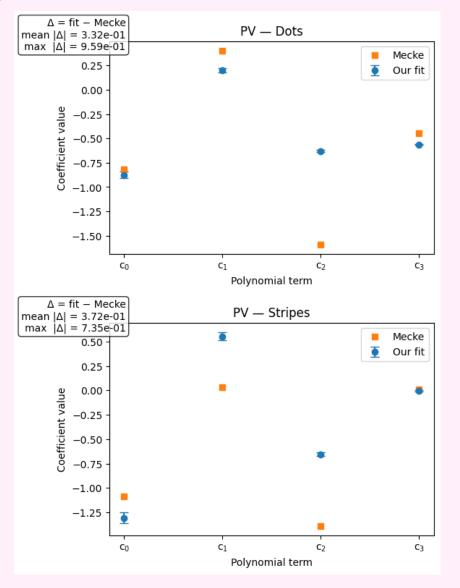




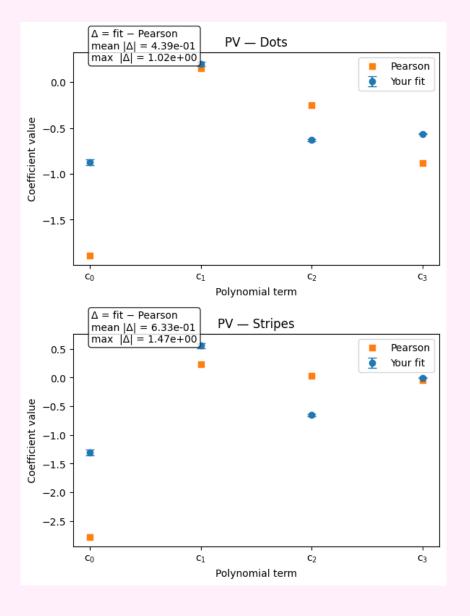
15	Dots	Stripes
æ	0.118	0.084
S	Dots	Stripes
æ	0.131	0.720
S	Dots	Stripes
A Y	0.176	0.137

Orientation agreement		
Same-sign	Pearson corr. coeff.	
1.000	-0.510	
0.278	-0.390	
0.764	0.151	

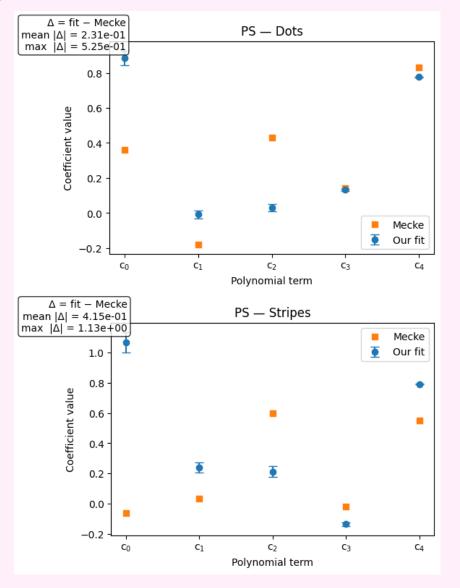
Polynomial coefficient errors



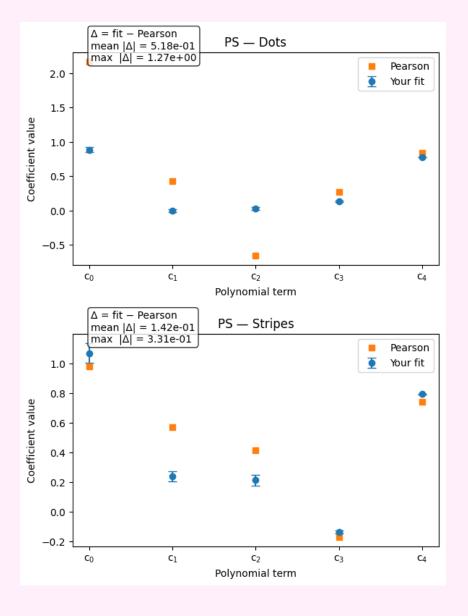




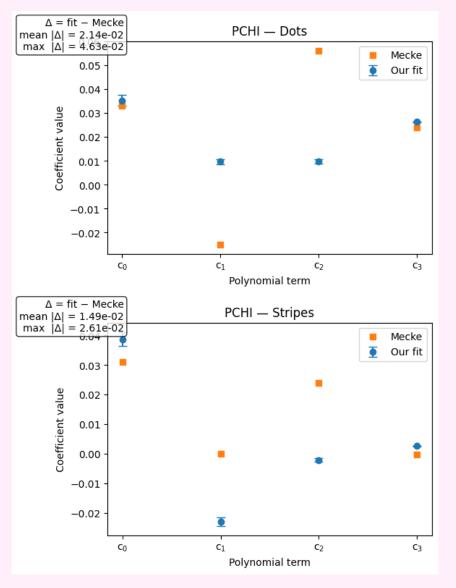
Polynomial coefficient errors



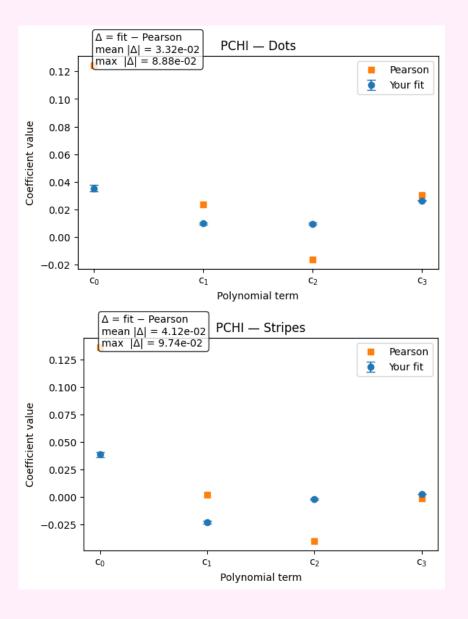




Polynomial coefficient errors







Extra slides: All Pearson plots

