

# Stochastic Turing Patterns of Trichomes in *Arabidopsis* Leaves

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**Young researchers @DMI**

V Workshop of the Department of Mathematics and Computer Science

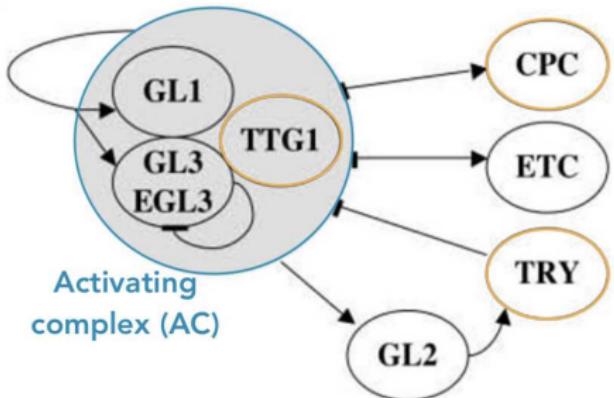
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DIPARTIMENTO  
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# Patterns in *Arabidopsis* leaves



- ▶ *A. thaliana* is a popular model organism in plant biology and genetics.

- ▶ Trichomes are epidermal hairs in the aerial parts of plants that

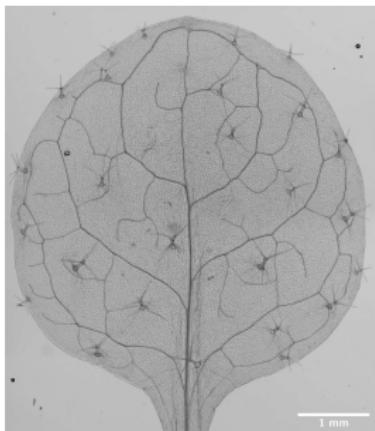
- ▶ provide a physical/chemical barrier against insect herbivores and UV light
- ▶ reduce transpiration
- ▶ increase tolerance to freezing

Commitment to trichome correlates with the accumulation of AC

# Experimental data set

- ▶ 14-day old wild-type plant leaf
- ▶ 6 biological replicate – 250 leaves

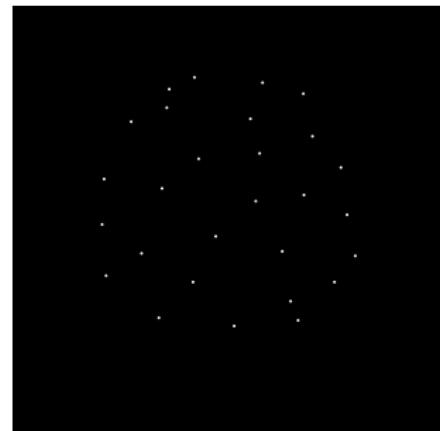
Transmission light



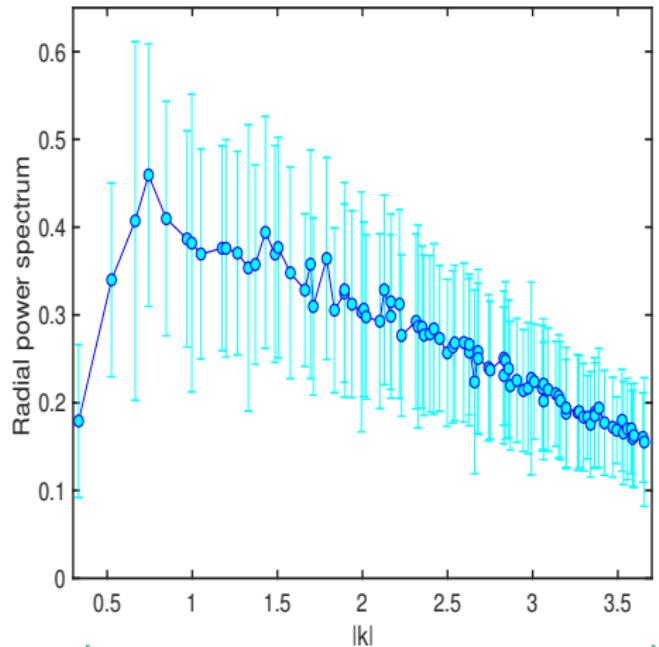
Cross polarizers



Positions of trichomes



# Experimental power spectrum

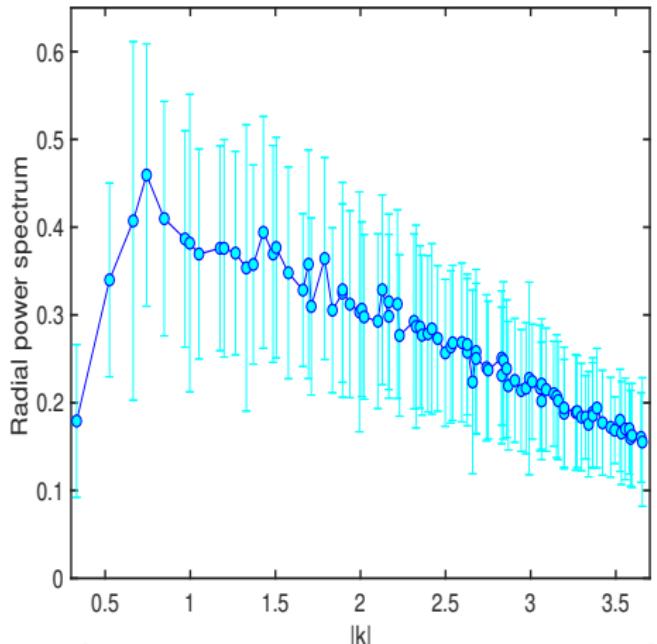


$\downarrow$   
 $|\mathbf{k}| \simeq 0.3$   
21 cells

$\downarrow$   
 $|\mathbf{k}| \simeq 3.5$   
cell radius

# Experimental power spectrum

How do patterns originate?

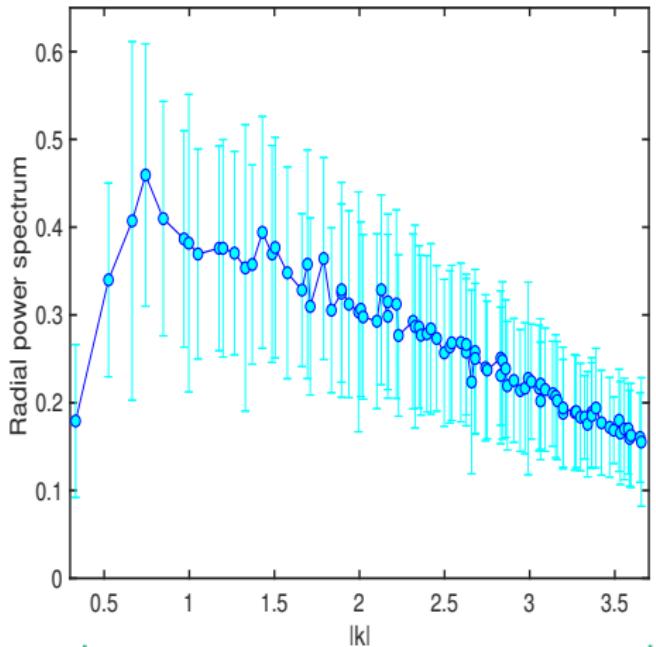


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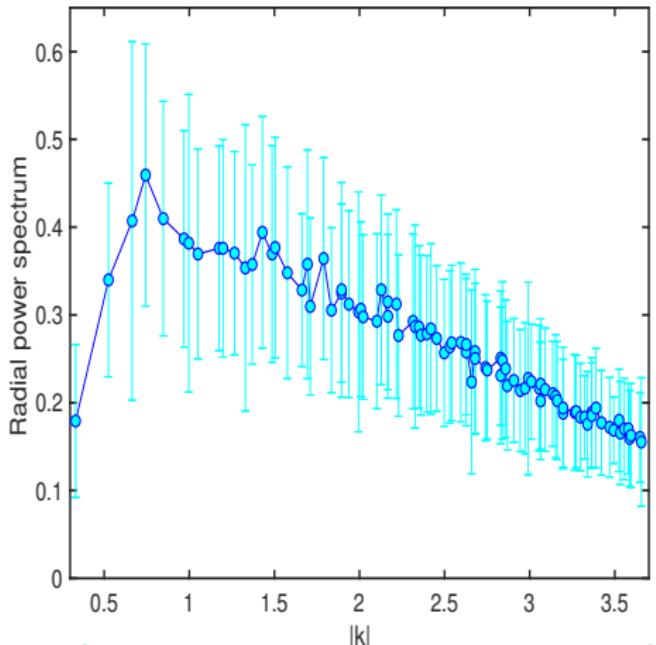
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### Deterministic patterns

- ▶ reaction-diffusion PDE
- ▶ Turing patterns:  
non-homogeneous  
perturbations of a stable  
steady state
- ▶ Turing patterns affected  
by random external noise

# Experimental power spectrum

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### Deterministic patterns

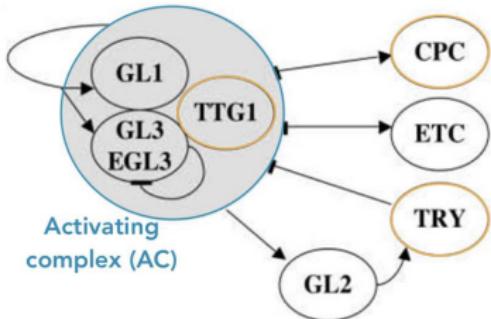
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### Stochastic Turing patterns

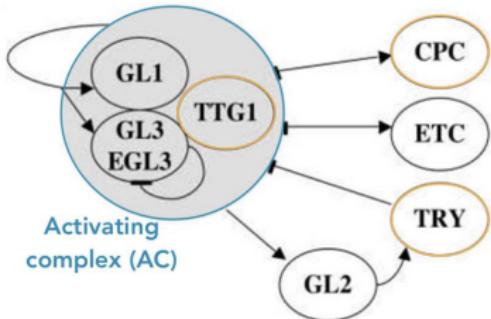
- ▶ Patterns induced by  
intrinsic noise

$|\mathbf{k}| \approx 3.5$   
cell radius

# Deterministic model

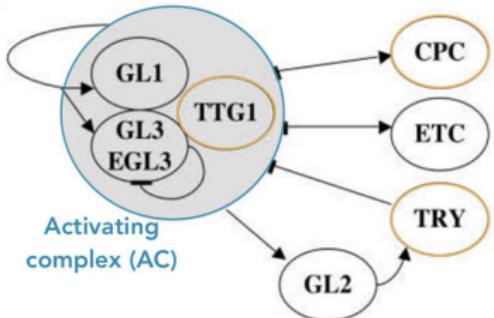



# Deterministic model



0	1	...	
		...	N-1

# Deterministic model

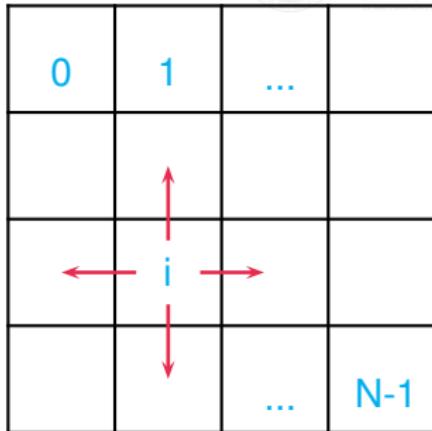
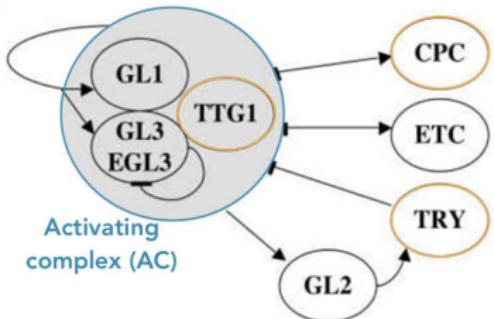


$(\phi_i^{AC}, \phi_i^I)$

0	1	...	
		...	N-1

$$\left\{ \begin{array}{l} \frac{d}{d\tau} \phi_i^{AC} = \alpha_{AC} + \beta_{AC} \frac{(\phi_i^{AC})^2}{\kappa + (\phi_i^{AC})^2} - \gamma \phi_i^{AC} \phi_i^I - \delta \phi_i^{AC} \\ \frac{d}{d\tau} \phi_i^I = \alpha_I + \beta_I \frac{(\phi_i^{AC})^2}{\kappa + (\phi_i^{AC})^2} - \gamma \phi_i^{AC} \phi_i^I - \delta \phi_i^I \end{array} \right.$$

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where  $\Delta_{ij} = A_{ij} - k_i \delta_{ij}$  is the Laplacian matrix,  $A_{ij}$  the adjacency matrix.

# Homogeneous stable equilibrium point

$$(\phi_{AC}^*, \phi_I^*)$$

## Rescaling

$$\tilde{\phi}_i^{AC} = \frac{\phi_i^{AC}}{\phi_{AC}^*} \quad \tilde{\phi}_i^I = \frac{\phi_i^I}{\phi_I^*} \quad \tilde{\alpha}_{AC} = \frac{\alpha_{AC}}{\phi_{AC}^* \cdot \delta} \quad \tilde{\alpha}_I = \frac{\alpha_I}{\phi_I^* \cdot \delta} \quad \tilde{\mu}_{AC} = \frac{\mu_{AC}}{\delta} \quad \tilde{\mu}_I = \frac{\mu_I}{\delta}$$

$$\tilde{\beta}_{AC} = \frac{\beta_{AC}}{\phi_{AC}^* \cdot \delta} \quad \tilde{\beta}_I = \frac{\beta_I}{\phi_I^* \cdot \delta} \quad \tilde{\gamma}_{AC} = \frac{\gamma \phi_I^*}{\delta} \quad \tilde{\gamma}_I = \frac{\gamma \phi_{AC}^*}{\delta} \quad \tilde{\kappa} = \frac{\kappa}{(\phi_{AC}^*)^2} \quad \tilde{\tau} = \tau \delta$$

$$\begin{cases} \frac{d}{d\tilde{\tau}} \tilde{\phi}_i^{AC} = & \tilde{\alpha}_{AC} + \tilde{\beta}_{AC} \frac{(\tilde{\phi}_i^{AC})^2}{\tilde{\kappa} + (\tilde{\phi}_i^{AC})^2} - \tilde{\gamma}_{AC} \tilde{\phi}_i^{AC} \tilde{\phi}_i^I - \tilde{\phi}_i^{AC} + \tilde{\mu}_{AC} \sum_{j=0}^{N-1} \Delta_{ij} \tilde{\phi}_j^{AC} \\ \frac{d}{d\tilde{\tau}} \tilde{\phi}_i^I = & \tilde{\alpha}_I + \tilde{\beta}_I \frac{(\tilde{\phi}_i^I)^2}{\tilde{\kappa} + (\tilde{\phi}_i^I)^2} - \tilde{\gamma}_I \tilde{\phi}_i^I \tilde{\phi}_i^{AC} - \tilde{\phi}_i^I + \tilde{\mu}_I \sum_{j=0}^{N-1} \Delta_{ij} \tilde{\phi}_j^I \end{cases}$$



# Linear stability analysis (Turing 1952)

- ▶ Introduce small **inhomogeneous** perturbations  $\delta\tilde{\phi}_i^{AC}$  and  $\delta\tilde{\phi}_i^I$ , to the uniform steady state as

$$(\tilde{\phi}_i^{AC}, \tilde{\phi}_i^I) = (\tilde{\phi}_{AC}^*, \tilde{\phi}_I^*) + (\delta\tilde{\phi}_i^{AC}, \delta\tilde{\phi}_i^I)$$

- ▶ Perform a Taylor expansion of the system
- ▶ Expand the non-homogeneous perturbations  $\delta\tilde{\phi}_i^{AC}$  and  $\delta\tilde{\phi}_i^I$  as

$$\delta\tilde{\phi}_i^{AC} = \sum_{\alpha=0}^{N-1} a_\alpha e^{\lambda^{(\alpha)} t} v_i^{(\alpha)} \quad \delta\tilde{\phi}_i^I = \sum_{\alpha=0}^{N-1} b_\alpha e^{\lambda^{(\alpha)} t} v_i^{(\alpha)}$$

where  $a_\alpha$  and  $b_\alpha$  can be self-consistently calculated, while  $\Delta\mathbf{v}^{(\alpha)} = \Lambda^{(\alpha)}\mathbf{v}^{(\alpha)}$  for  $\alpha = 0, \dots, N-1$ .

## Dispersion relation

$$\det(\mathbf{J} + \mathbf{D}\Lambda^{(\alpha)} - \lambda_\alpha \mathbb{I}_2) = 0$$

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$$\det(\mathbf{J} + D\Lambda^{(\alpha)} - \lambda_\alpha \mathbb{I}_2) = 0$$

- ▶  $\mathbf{J}$  is the Jacobian matrix
- ▶  $D = \begin{pmatrix} \tilde{\mu}_{AC} & 0 \\ 0 & \tilde{\mu}_I \end{pmatrix}$
- ▶  $\Lambda^{(\alpha)} \longleftrightarrow -|\mathbf{k}|^2$

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Imposed perturbations get magnified if  $Re(\lambda) > 0$

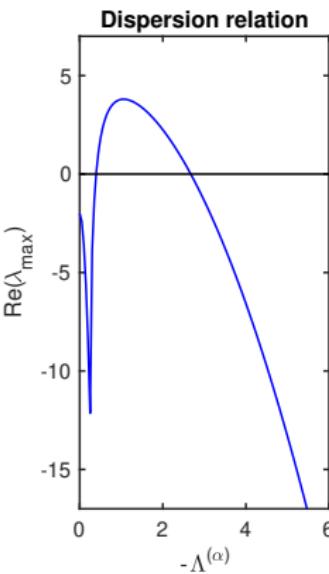
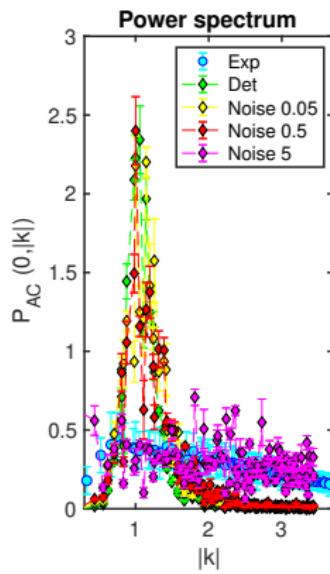
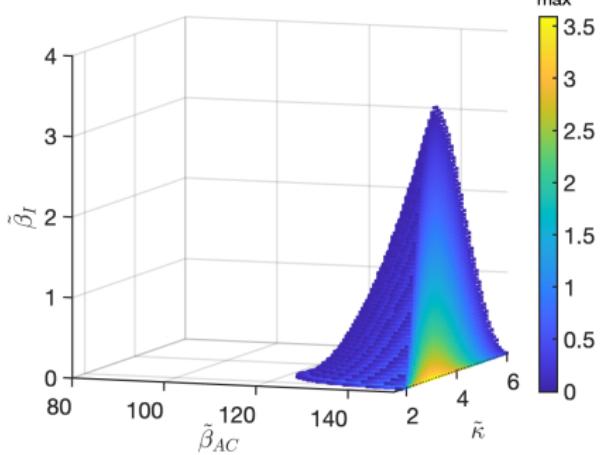
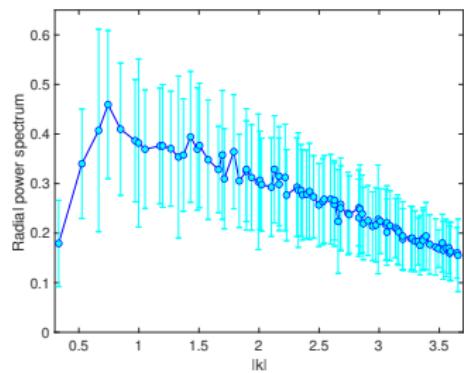
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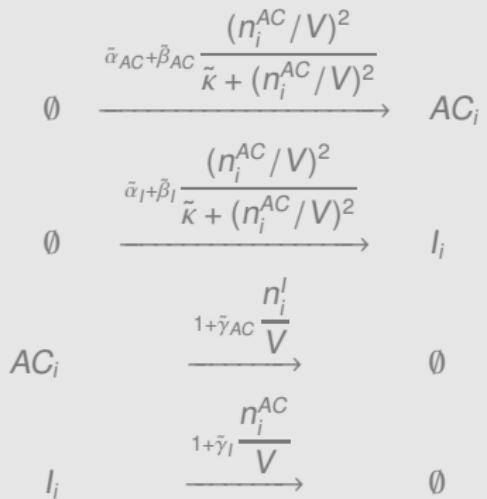
# Deterministic Turing instability region



# Stochastic model

$V$  is the volume of individual cells

## Reaction rules



## Transition rates

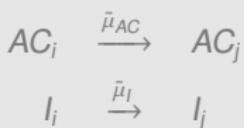
$$T_1(n_i^{AC} + 1 | n_i^{AC}) = \tilde{\alpha}_{AC} + \tilde{\beta}_{AC} \frac{(n_i^{AC}/V)^2}{\tilde{\kappa} + (n_i^{AC}/V)^2}$$

$$T_2(n_i^I + 1 | n_i^I) = \tilde{\alpha}_I + \tilde{\beta}_I \frac{(n_i^{AC}/V)^2}{\tilde{\kappa} + (n_i^{AC}/V)^2}$$

$$T_3(n_i^{AC} - 1 | n_i^{AC}) = \left(1 + \tilde{\gamma}_{AC} \frac{n_i^I}{V}\right) \frac{n_i^{AC}}{V}$$

$$T_4(n_i^I - 1 | n_i^I) = \left(1 + \tilde{\gamma}_I \frac{n_i^{AC}}{V}\right) \frac{n_i^I}{V}$$

## Diffusion rules



$$T_5(n_i^{AC} - 1, n_j^{AC} + 1 | n_i^{AC}, n_j^{AC}) = \tilde{\mu}_{AC} \frac{n_i^{AC}}{V}$$

$$T_6(n_i^I - 1, n_j^I + 1 | n_i^I, n_j^I) = \tilde{\mu}_I \frac{n_i^I}{V}$$

# Master equation

## State of the system

$$(\mathbf{n}^{AC}, \mathbf{n}^I) \quad \text{with} \quad \begin{aligned} \mathbf{n}^{AC} &= (n_0^{AC}, \dots, n_{N-1}^{AC}) \\ \mathbf{n}^I &= (n_0^I, \dots, n_{N-1}^I) \end{aligned}$$

$$\begin{aligned} \frac{d}{dt} P(\mathbf{n}^{AC}, \mathbf{n}^I; t) = & \sum_{i=0}^{N-1} \left\{ (\epsilon_{AC,i}^- - 1) \mathbb{T}_1(n_i^{AC} + 1 | n_i^{AC}) + (\epsilon_{I,i}^- - 1) \mathbb{T}_2(n_i^I + 1 | n_i^I) \right. \\ & + (\epsilon_{AC,i}^+ - 1) \mathbb{T}_3(n_i^{AC} - 1 | n_i^{AC}) + (\epsilon_{I,i}^+ - 1) \mathbb{T}_4(n_i^I - 1 | n_i^I) \\ & + \sum_{j=0}^{N-1} A_{ij} \left[ (\epsilon_{AC,i}^+ \epsilon_{AC,j}^+ - 1) \mathbb{T}_5(n_i^{AC} - 1, n_j^{AC} + 1 | n_i^{AC}, n_j^{AC}) \right. \\ & \left. \left. + (\epsilon_{I,i}^+ \epsilon_{I,j}^+ - 1) \mathbb{T}_6(n_i^I - 1, n_j^I + 1 | n_i^I, n_j^I) \right] \right\} P(\mathbf{n}^{AC}, \mathbf{n}^I; t) \end{aligned}$$

## Step operators

$$\epsilon_{AC,i}^\pm f(\mathbf{n}^{AC}, \mathbf{n}^I) = f(\dots, n_i^{AC} \pm 1, \dots, \mathbf{n}^I)$$

$$\epsilon_{I,i}^\pm f(\mathbf{n}^{AC}, \mathbf{n}^I) = f(\mathbf{n}^{AC}, \dots, n_i^I \pm 1, \dots).$$

# The van Kampen expansion



## New variables

$$\frac{n_i^{AC}}{V} = \tilde{\phi}_i^{AC} + \frac{1}{\sqrt{V}} \xi_{AC,i}$$

$$\frac{n_i^I}{V} = \tilde{\phi}_i^I + \frac{1}{\sqrt{V}} \xi_{I,i}$$

# The van Kampen expansion



## New variables

## Deterministic variables

$$\phi_i = \lim_{V \rightarrow \infty} \frac{\langle n_i \rangle}{V}$$

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## Stochastic variables (fluctuations)

# The van Kampen expansion



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## Deterministic variables

$$\phi_i = \lim_{V \rightarrow \infty} \frac{\langle n_i \rangle}{V}$$

## New probability

$$P(n^{AC}, n^I; t) \rightarrow \Pi(\xi_{AC}, \xi_I; t)$$

## Stochastic variables (fluctuations)

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## Stochastic variables (fluctuations)

## System-size expansion w.r.t. $1/\sqrt{V}$

- $\frac{d}{dt} P(n^{AC}, n^I; t) = \sum_{i=0}^{N-1} \left( \frac{\partial \Pi}{\partial t} - \frac{\partial \Pi}{\partial \xi_{AC,i}} \sqrt{V} \dot{\tilde{\phi}}_i^{AC} - \frac{\partial \Pi}{\partial \xi_{I,i}} \sqrt{V} \dot{\tilde{\phi}}_i^I \right)$
- $\epsilon_{X,i}^\pm \simeq 1 \pm \frac{1}{\sqrt{V}} \frac{\partial}{\partial \xi_{X,i}} + \frac{1}{2V} \frac{\partial^2}{\partial \xi_{X,i}^2} \quad \text{for } X = AC, I$

# The leading order Order



Collecting together terms involving  $1/\sqrt{V}$  we get:

## Deterministic mean-field equations ( $\lim V \rightarrow \infty$ )

$$\begin{cases} \frac{d}{d\tilde{\tau}} \tilde{\phi}_i^{AC} &= \tilde{\alpha}_{AC} + \tilde{\beta}_{AC} \frac{(\tilde{\phi}_i^{AC})^2}{\tilde{\kappa} + (\tilde{\phi}_i^{AC})^2} - \tilde{\gamma}_{AC} \tilde{\phi}_i^{AC} \tilde{\phi}_i^I - \tilde{\phi}_i^{AC} + \tilde{\mu}_{AC} \sum_{j=0}^{N-1} \Delta_{ij} \tilde{\phi}_j^{AC} \\ \frac{d}{d\tilde{\tau}} \tilde{\phi}_i^I &= \tilde{\alpha}_I + \tilde{\beta}_I \frac{(\tilde{\phi}_i^{AC})^2}{\tilde{\kappa} + (\tilde{\phi}_i^{AC})^2} - \tilde{\gamma}_I \tilde{\phi}_i^{AC} \tilde{\phi}_i^I - \tilde{\phi}_i^I + \tilde{\mu}_I \sum_{j=0}^{N-1} \Delta_{ij} \tilde{\phi}_j^I \end{cases}$$

# Next-to-leading order

## Fokker-Planck equation

$$\frac{\partial}{\partial \tau} \Pi = \sum_{i=0}^{N-1} \left( - \sum_{q=1}^2 \frac{\partial}{\partial \xi_{q,i}} (A_{q,i} \Pi) + \frac{1}{2} \sum_{q,l=1}^2 \sum_{j=0}^{N-1} (B_{ql,ij} \Pi) \right)$$

with

$$A_{q,i} = \sum_{l=1}^2 \sum_{j=0}^{N-1} M_{ql,ij} \xi_{s,j}$$

where the  $2N \times 2N$  matrices **M** and **B** are given by

$$M_{ql,ij} = M_{ql}^{(NS)} \delta_{ij} + M_{ql}^{(SP)} \Delta_{ij}$$

$$B_{ql,ij} = B_{ql}^{(NS)} \delta_{ij} + B_{ql}^{(SP)} \Delta_{ij}$$

and

$$q = 1 \equiv AC$$

$$q = 2 \equiv I$$

$$\mathbf{M}^{(NS)} = \begin{pmatrix} \frac{2\tilde{\beta}_{AC}\tilde{\kappa}}{(\tilde{\kappa}+1)^2} - \tilde{\gamma}_{AC} - 1 & -\tilde{\gamma}_{AC} \\ \frac{2\tilde{\beta}_I\tilde{\kappa}}{(\tilde{\kappa}+1)^2} - \tilde{\gamma}_I & -\tilde{\gamma}_I - 1 \end{pmatrix} \quad \mathbf{M}^{(SP)} = \begin{pmatrix} \tilde{\mu}_{AC} & 0 \\ 0 & \tilde{\mu}_I \end{pmatrix}$$

$$\mathbf{B}^{(NS)} = \mathbf{S} \cdot \begin{pmatrix} \tilde{\alpha}_{AC} + \frac{\tilde{\beta}_{AC}}{\tilde{\kappa}+1} & 0 & 0 & 0 \\ 0 & \tilde{\alpha}_I + \frac{\tilde{\beta}_I}{\tilde{\kappa}+1} & 0 & 0 \\ 0 & 0 & 1 + \tilde{\gamma}_{AC} & 0 \\ 0 & 0 & 0 & 1 + \tilde{\gamma}_I \end{pmatrix} \mathbf{S}^t$$

$$\mathbf{B}^{(SP)} = \begin{pmatrix} -2\tilde{\mu}_{AC} & 0 \\ 0 & -2\tilde{\mu}_I \end{pmatrix} .$$

The matrix  $\mathbf{S}$  is the stoichiometric matrix that reflects the local reaction rules and takes the form

$$\mathbf{S} = \begin{pmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{pmatrix}$$

## Langevin equation

$$\frac{d\xi_{q,i}}{d\tau} = \sum_{l=1}^2 \sum_{j=0}^{N-1} M_{ql,ij} \xi_{l,j} + \eta_{q,i}$$

where  $\eta_{q,i}$  is a Gaussian white noise, with zero mean and correlator

$$\langle \eta_{q,i}(\tau) \eta_{l,j}(\tau') \rangle = B_{ql,ij} \delta(\tau - \tau')$$

## Temporal and spatially-discrete Fourier transform

$$\hat{f}_\alpha(\omega) = \int_0^{+\infty} d\tau \sum_{j=0}^{N-1} f_j(\tau) v_j^{(\alpha)} e^{i\omega\tau}$$

# Power spectrum

The application of the Fourier transform to the Langevin equation gives

$$\hat{\xi}_{q,\alpha} = \sum_{l=1}^2 \mathbf{F}_{ql}^{-1} \hat{\eta}_{l,\alpha}$$

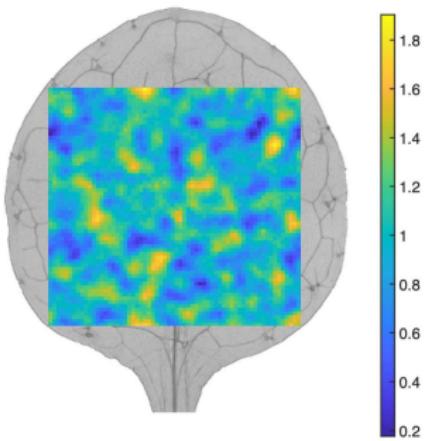
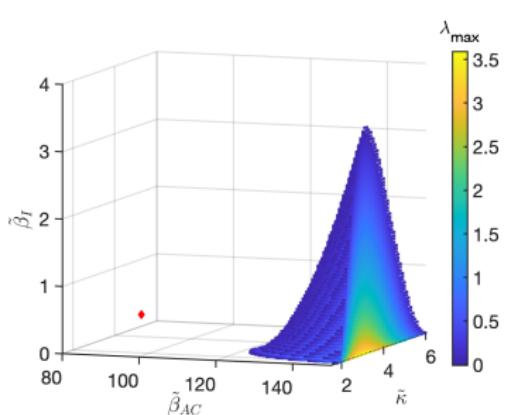
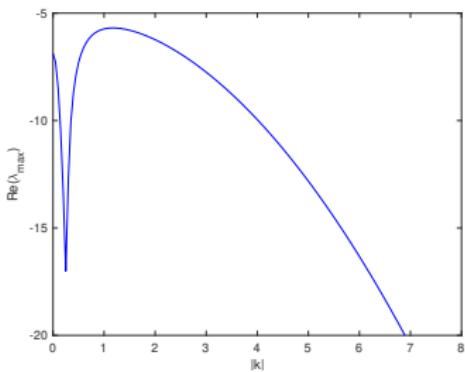
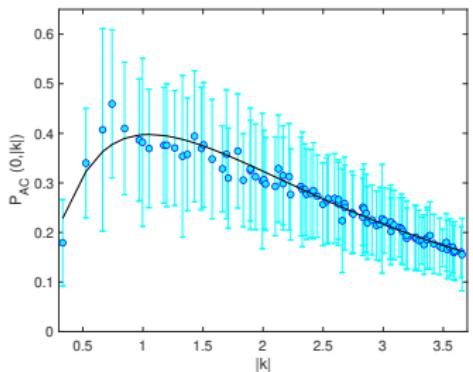
where  $\mathbf{F} = (-i\omega\mathbb{I} - M^{(NS)} - M^{(SP)}\Lambda^{(\alpha)})$ .

## Power spectrum

$$P_q(\omega, \Lambda^{(\alpha)}) = \langle |\hat{\xi}_{q,\alpha}(\omega)|^2 \rangle = \sum_{l,m=1}^2 \mathbf{F}_{ql}^{-1} (B_{lm}^{(NS)} + B_{lm}^{(SP)} \Lambda^{(\alpha)}) (\mathbf{F}^\dagger)_{mq}^{-1}$$

where the symbol  $\dagger$  denotes the adjoint operator.

# Experiment – theory – simulations



# Co-authors

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**Thanks for your attention!**