

September the 29th, 2015, Palermo



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2015

University of Palermo - September 28 - October 2, 2015 - Conference Chairs: Ezio Puppin (CNISM) - Corrado Spinella (CNR)

Timoteo Carletti

Turing patterns in multiplex networks



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Acknowledgements

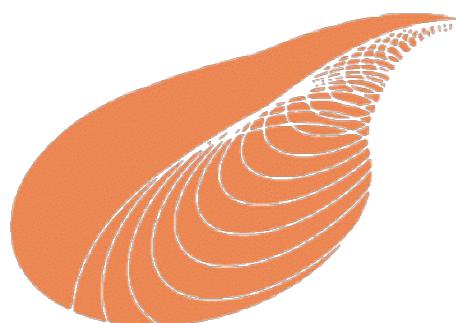
M. Asllani, University of Namur (Belgium)

D.M. Busiello, University of Padova (Italy)

D. Fanelli, University of Firenze (Italy)

J. Petit, University of Namur (Belgium)

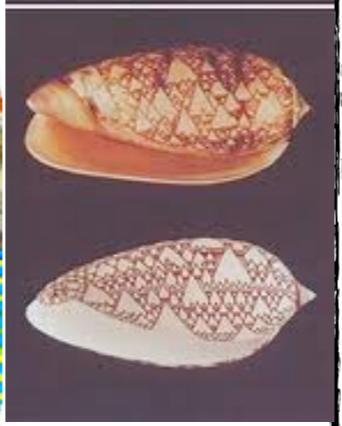
G. Planchon, University of Namur (Belgium)



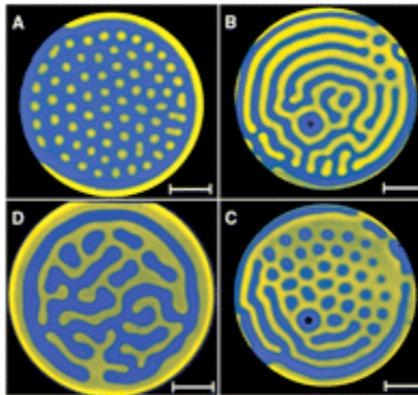
IAP VII/19 - DYSKO



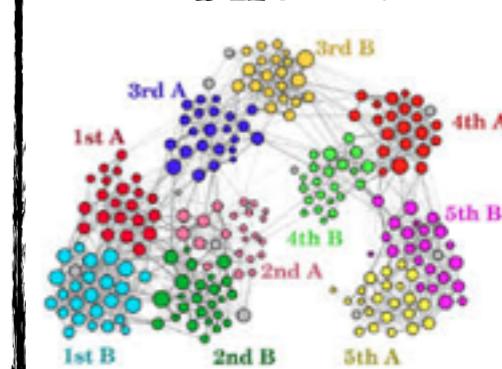
Patterns are ubiquitous



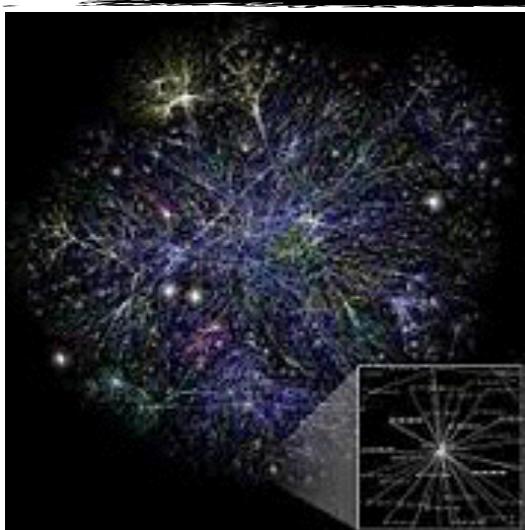
Animal kingdom



Chemistry



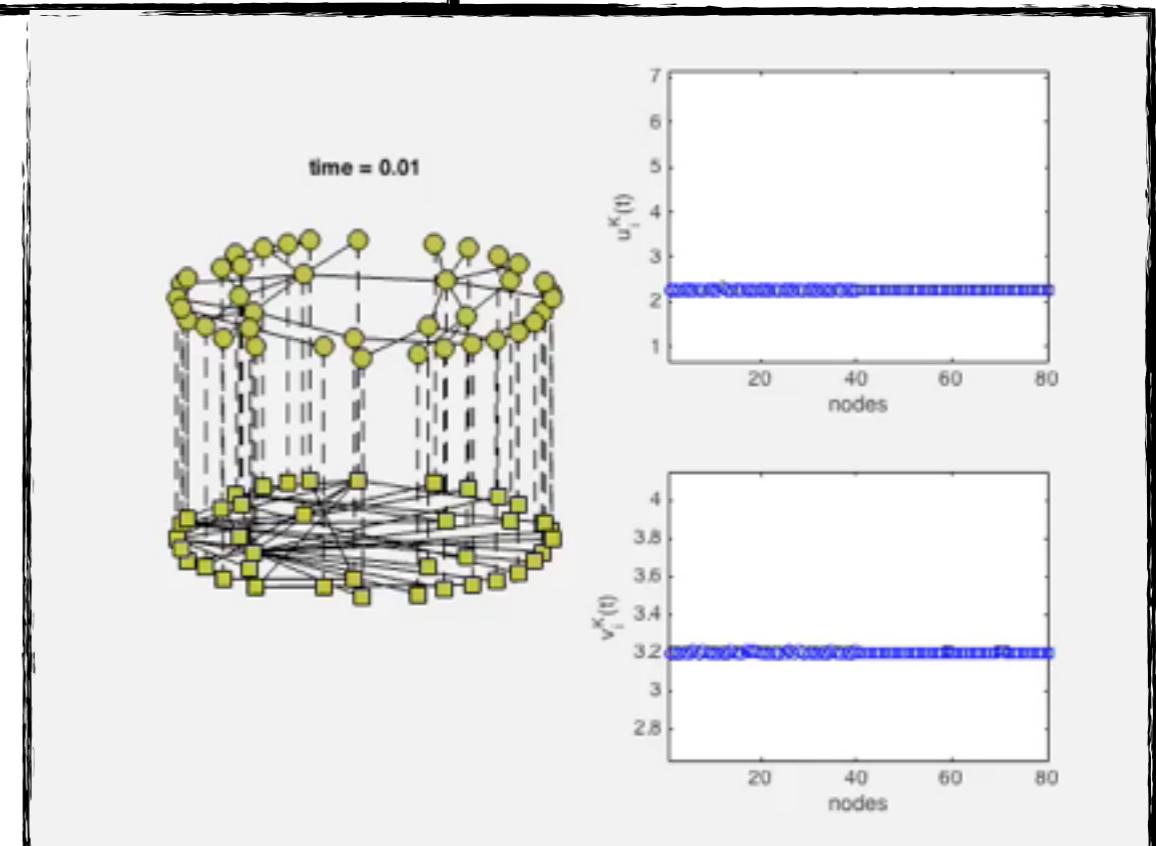
SocioPatterns



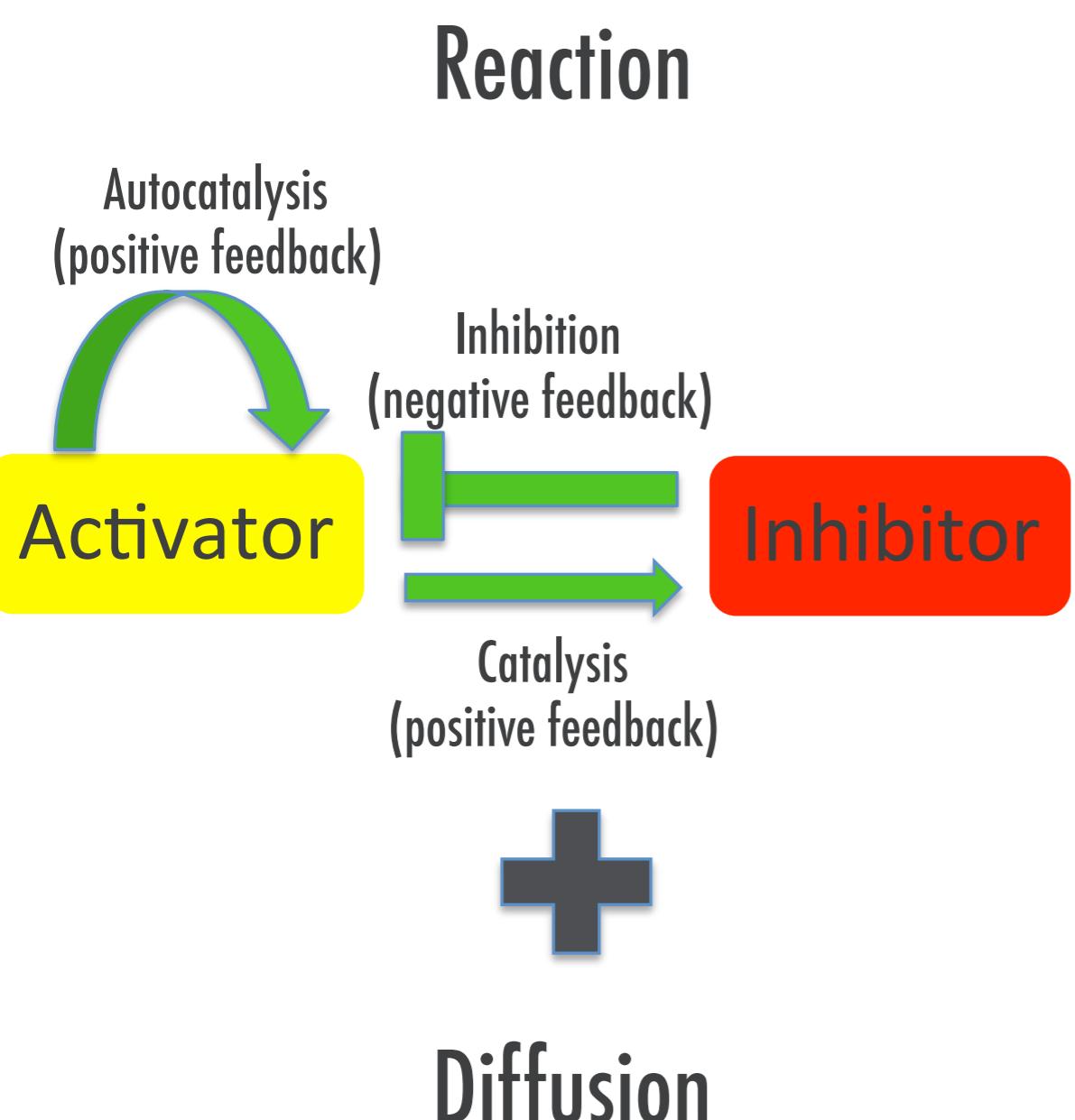
Internet



Twitter



One possible mechanism: Turing instability



$u(x, y, t)$: Amount of activator at time t and position (x, y)

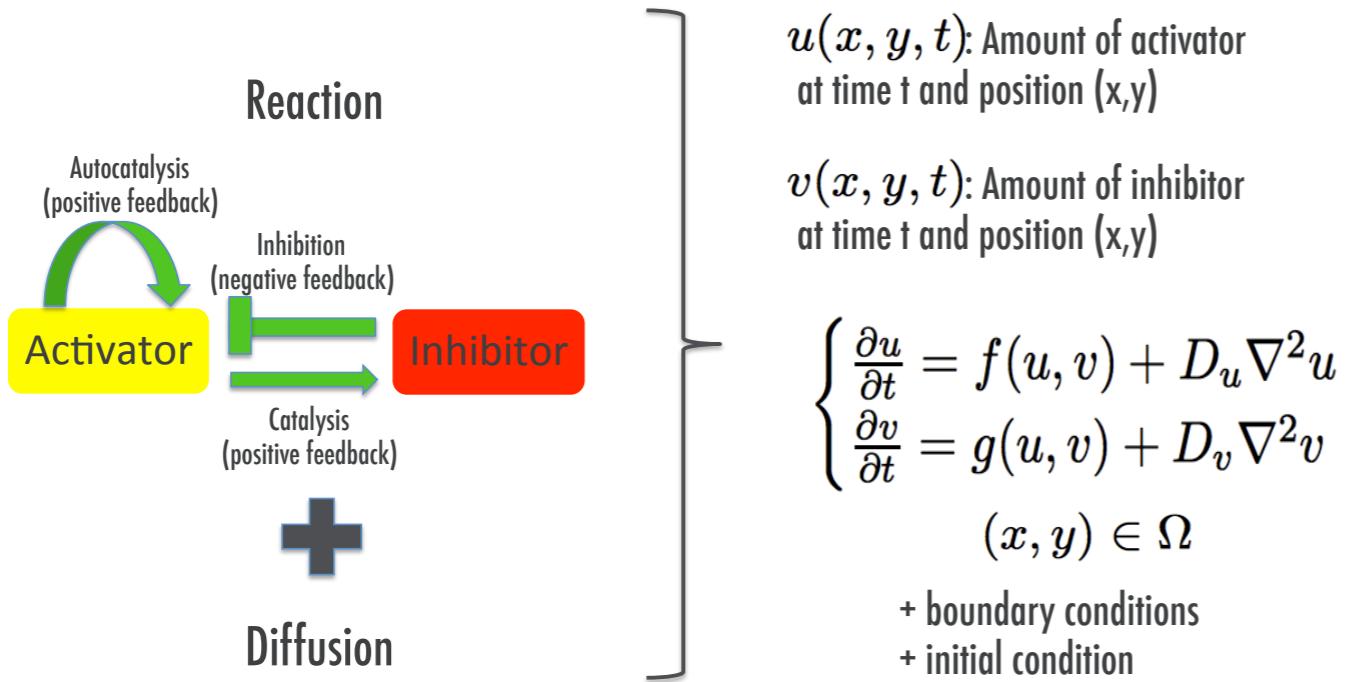
$v(x, y, t)$: Amount of inhibitor at time t and position (x, y)

$$\begin{cases} \frac{\partial u}{\partial t} = f(u, v) + D_u \nabla^2 u \\ \frac{\partial v}{\partial t} = g(u, v) + D_v \nabla^2 v \end{cases}$$

$$(x, y) \in \Omega$$

+ boundary conditions
+ initial condition

One possible mechanism: Turing instability



Diffusion can drive an instability by perturbing a homogeneous stable fixed point. Hence as the perturbation grows, non linearities enter into the game yielding an asymptotic, spatially inhomogeneous, steady state (stationary pattern).

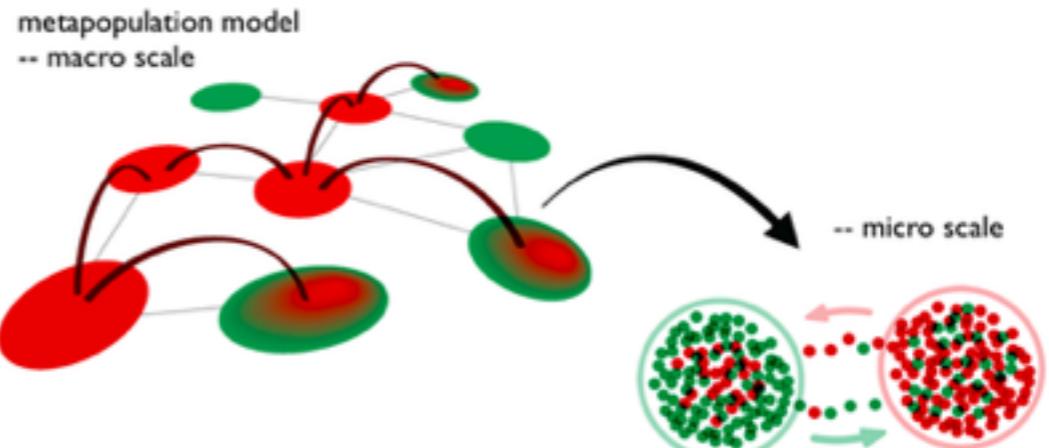
A.M.Turing, *The chemical basis of morphogenesis*, Phil. Trans. R Soc London B, 237, (1952), pp.37

Extension to networks



world flights map

Reactions occur at each node.
Diffusion occurs across edges.



Metapopulation models

e.g. in the framework of ecology:

May R., *Will a large complex system be stable?* Nature, 238, pp. 413, (1972)

Patterns : sets of nodes whose asymptotic state is far from the homogeneous equilibrium.

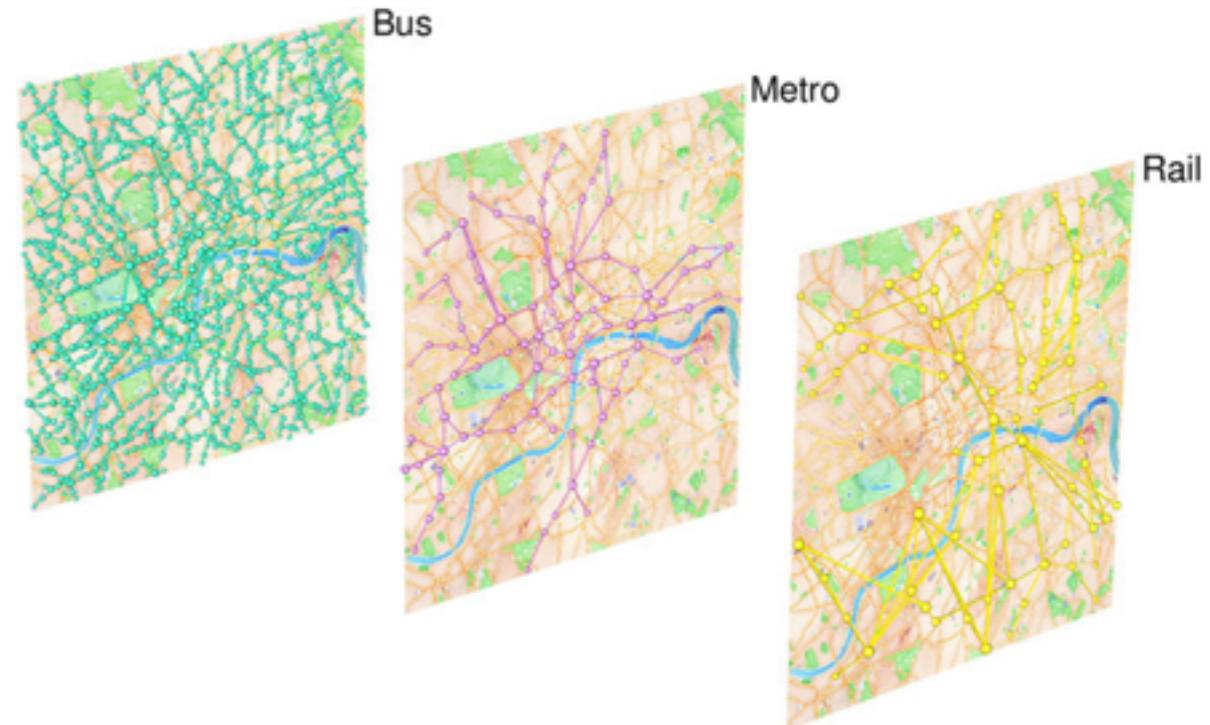
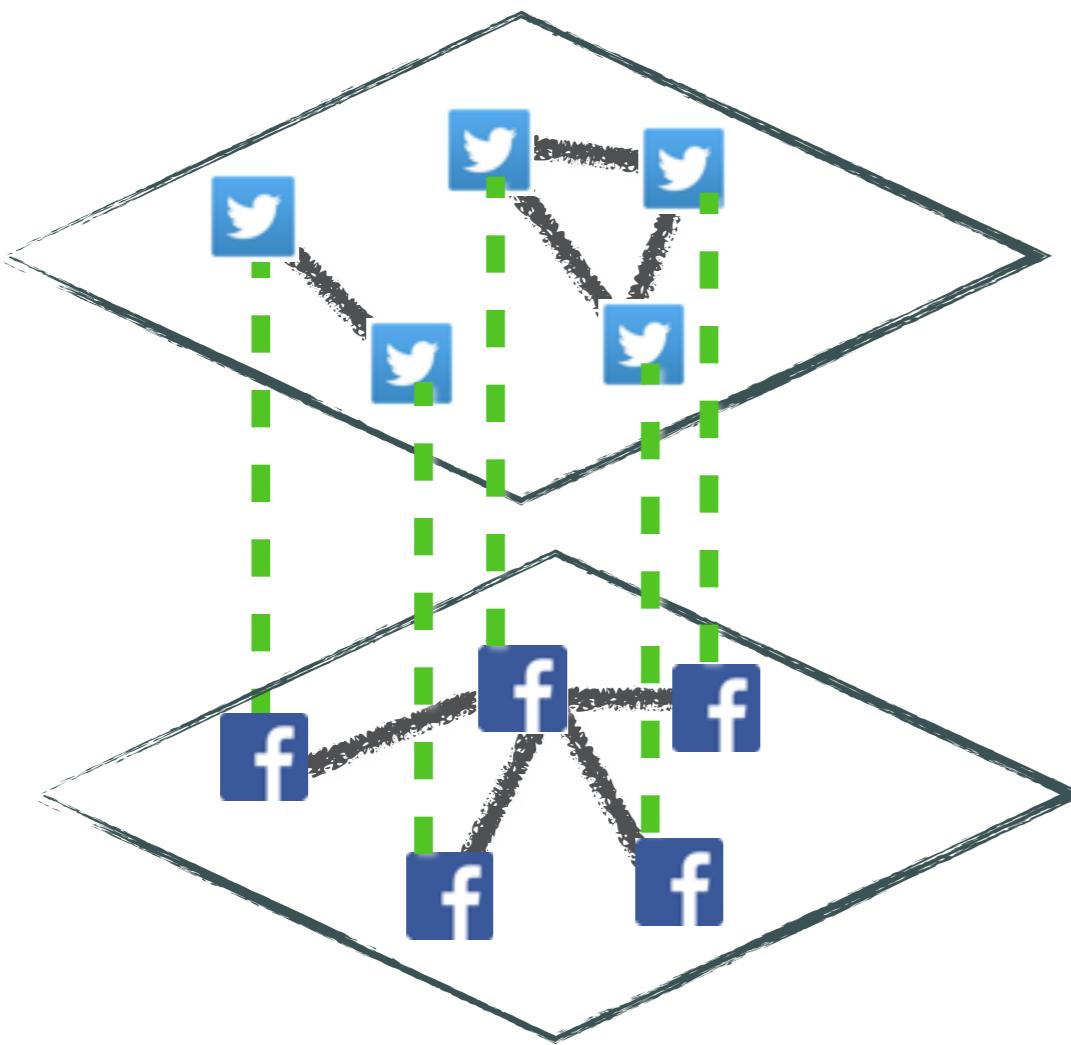
Nakao H. and Mikhailov A. S., *Turing patterns in network-organized activator-inhibitor systems*, Nat. Phys., 6, pp. 544 (2010)

Systems composed by layers of networks: Multiplexes

Social networks

layers=different social networks

nodes=same agent in each SN



Transportation networks

layers=different modalities

nodes=same spatial location

Turing instabilities on multiplex networks

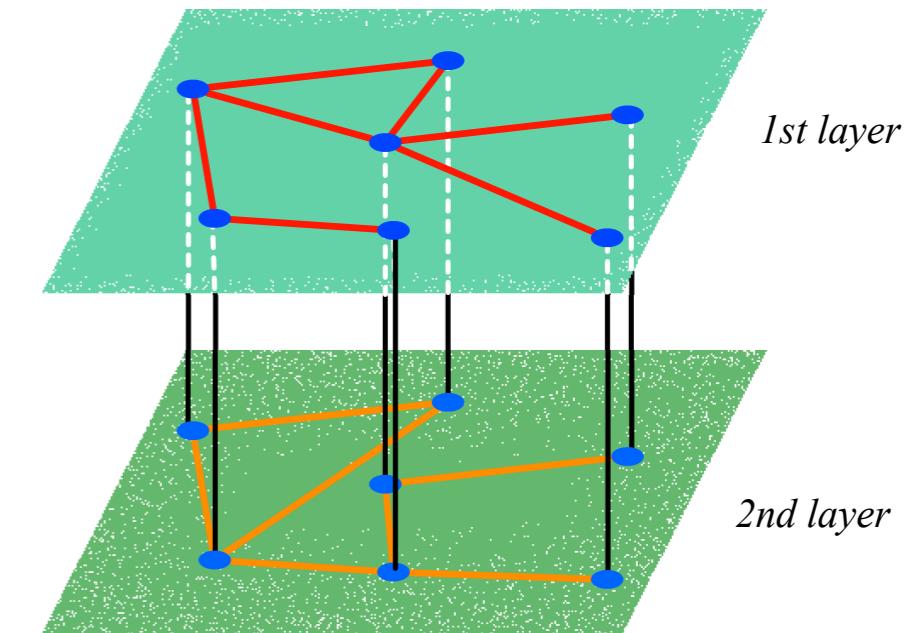
With $K=1,2$ ($K=3$ should be read $K=1$)

adjacency matrix of
layer K

$$L_{ij}^K = A_{ij}^K - \delta_{ij} k_i^K$$

degree of ith note
in layer K

Laplacian matrix of
layer K



The same Ω nodes are present in each layer

$D_{u,v}^K$ **inter-layer diffusion coefficient**

$D_{u,v}^{12}$ **intra-layer diffusion coefficient**

$$\begin{cases} \dot{u}_i^K &= f(u_i^K, v_i^K) + D_u^K \sum_{j=1}^{\Omega} L_{ij}^K u_j^K + D_u^{12} (u_i^{K+1} - u_i^K) \\ \dot{v}_i^K &= g(u_i^K, v_i^K) + D_v^K \sum_{j=1}^{\Omega} L_{ij}^K v_j^K + D_v^{12} (v_i^{K+1} - v_i^K) \end{cases}$$

General strategy

- 1) Determine an homogeneous solution $(u_j^K, v_j^K) = (\hat{u}, \hat{v})$
(assumed to be stable when $D_u^K = D_v^K = D_{u,v}^{12} = 0$)

General strategy

2) Linearize around this solution

$$u_j^K = \hat{u} + \delta u_j^K$$

$$v_j^K = \hat{v} + \delta v_j^K$$

$$\begin{pmatrix} \dot{\delta u} \\ \dot{\delta v} \end{pmatrix} = \tilde{\mathcal{J}} \begin{pmatrix} \delta u \\ \delta v \end{pmatrix}$$

$$\tilde{\mathcal{J}} = \begin{pmatrix} f_u \mathbf{I}_{2\Omega} + \mathcal{L}_u + D_u^{12} \mathcal{I} & f_v \mathbf{I}_{2\Omega} \\ g_u \mathbf{I}_{2\Omega} & g_v \mathbf{I}_{2\Omega} + \mathcal{L}_v + D_v^{12} \mathcal{I} \end{pmatrix}$$

Supra-Laplacian matrix $\mathcal{L}_u + D_u^{12} \mathcal{I}$

$$\mathcal{L}_u = \begin{pmatrix} D_u^1 \mathbf{L}^1 & \mathbf{0} \\ \mathbf{0} & D_u^2 \mathbf{L}^2 \end{pmatrix}$$

$$\mathcal{I} = \begin{pmatrix} -\mathbf{I}_\Omega & \mathbf{I}_\Omega \\ \mathbf{I}_\Omega & -\mathbf{I}_\Omega \end{pmatrix}$$

3) Study the spectrum of

$$\tilde{\mathcal{J}} = \begin{pmatrix} f_u \mathbf{I}_{2\Omega} + \mathcal{L}_u + D_u^{12} \mathcal{I} & f_v \mathbf{I}_{2\Omega} \\ g_u \mathbf{I}_{2\Omega} & g_v \mathbf{I}_{2\Omega} + \mathcal{L}_v + D_v^{12} \mathcal{I} \end{pmatrix}$$

to determine the existence of eigenvalues such that

$$\Re \lambda(D_{u,v}^{12}, D_{u,v}^K) > 0$$

Very hard for generic topologies

Small intra-layer diffusion case

Assume $D_v^{12} = \epsilon \ll 1$ $D_u^{12} = \mathcal{O}(\epsilon)$

$$\tilde{\mathcal{J}} = \begin{pmatrix} f_u \mathbf{I}_{2\Omega} + \mathcal{L}_u & f_v \mathbf{I}_{2\Omega} \\ g_u \mathbf{I}_{2\Omega} & g_v \mathbf{I}_{2\Omega} + \mathcal{L}_v \end{pmatrix} + \epsilon \begin{pmatrix} \frac{D_u^{12}}{D_v^{12}} L^1 & 0 \\ 0 & L^2 \end{pmatrix}$$
$$= \tilde{\mathcal{J}}_0 + \epsilon \mathcal{D}_0$$

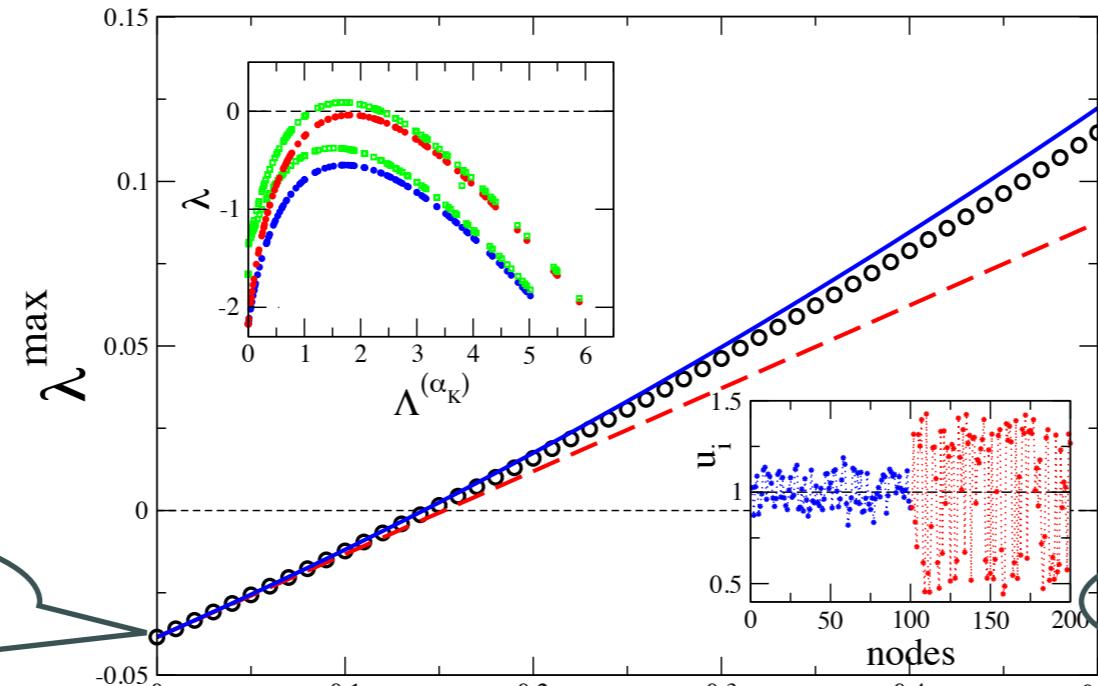
Perturbative approach to compute the spectrum

$$\lambda^{max}(\epsilon) = \lambda_0^{max} + \epsilon (U_0 \mathcal{D}_0 V_0)_{k_{max} k_{max}} + \mathcal{O}(\epsilon^2)$$

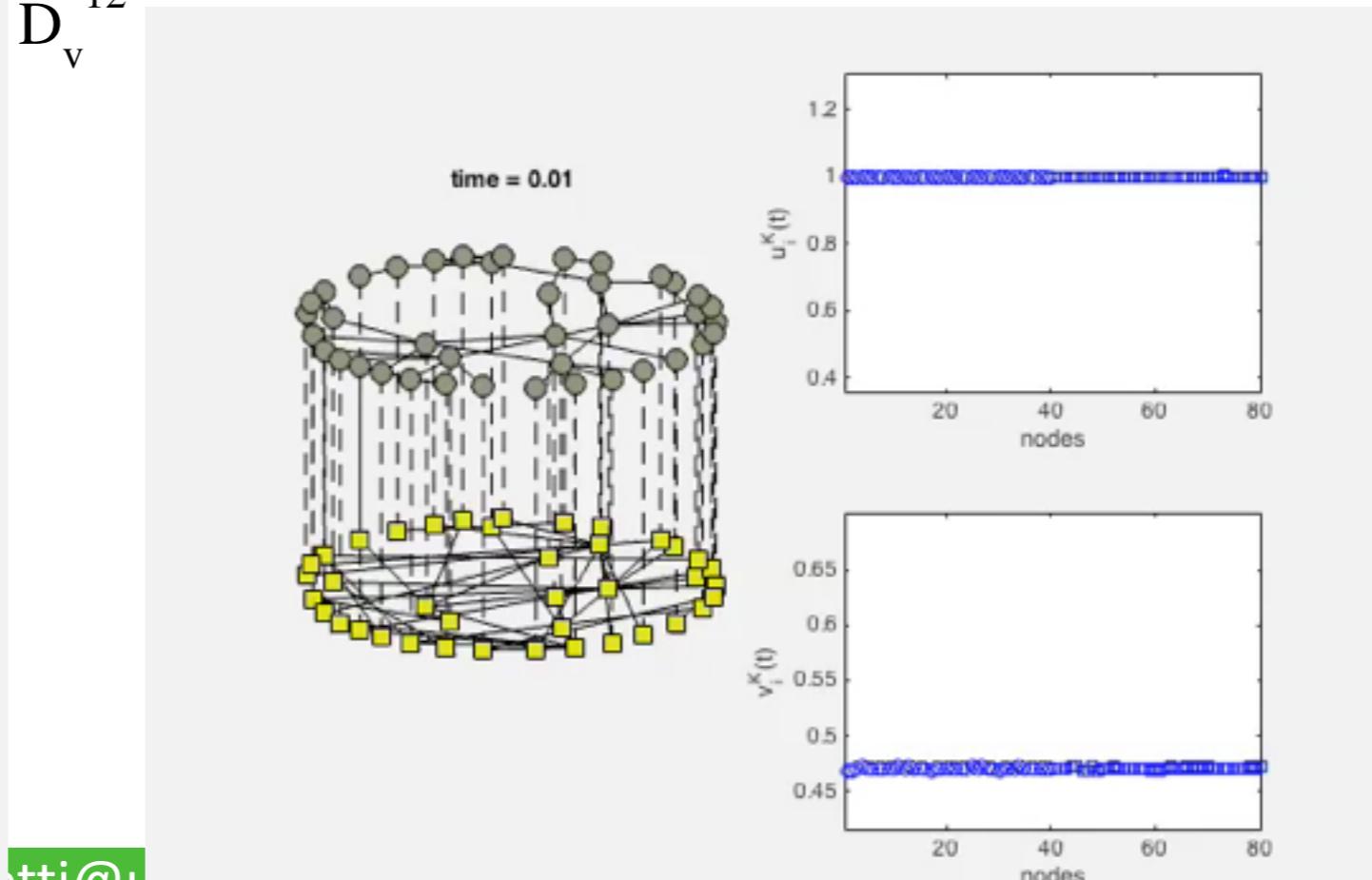
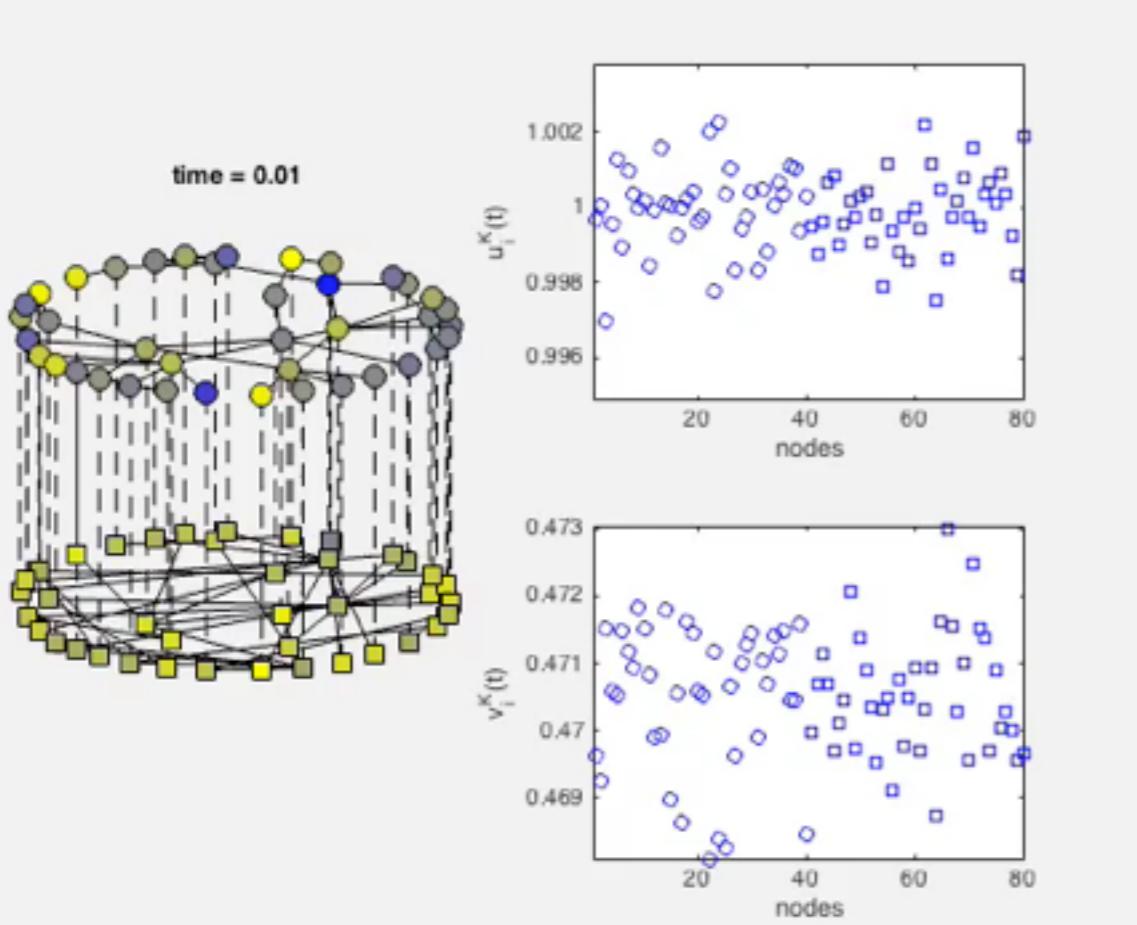
$$\lambda_0^{max} = \max \lambda_k(\epsilon = 0) \quad k_{max} = \arg \max \lambda_k(\epsilon = 0)$$

Small intra-layer diffusion case: onset of patterns

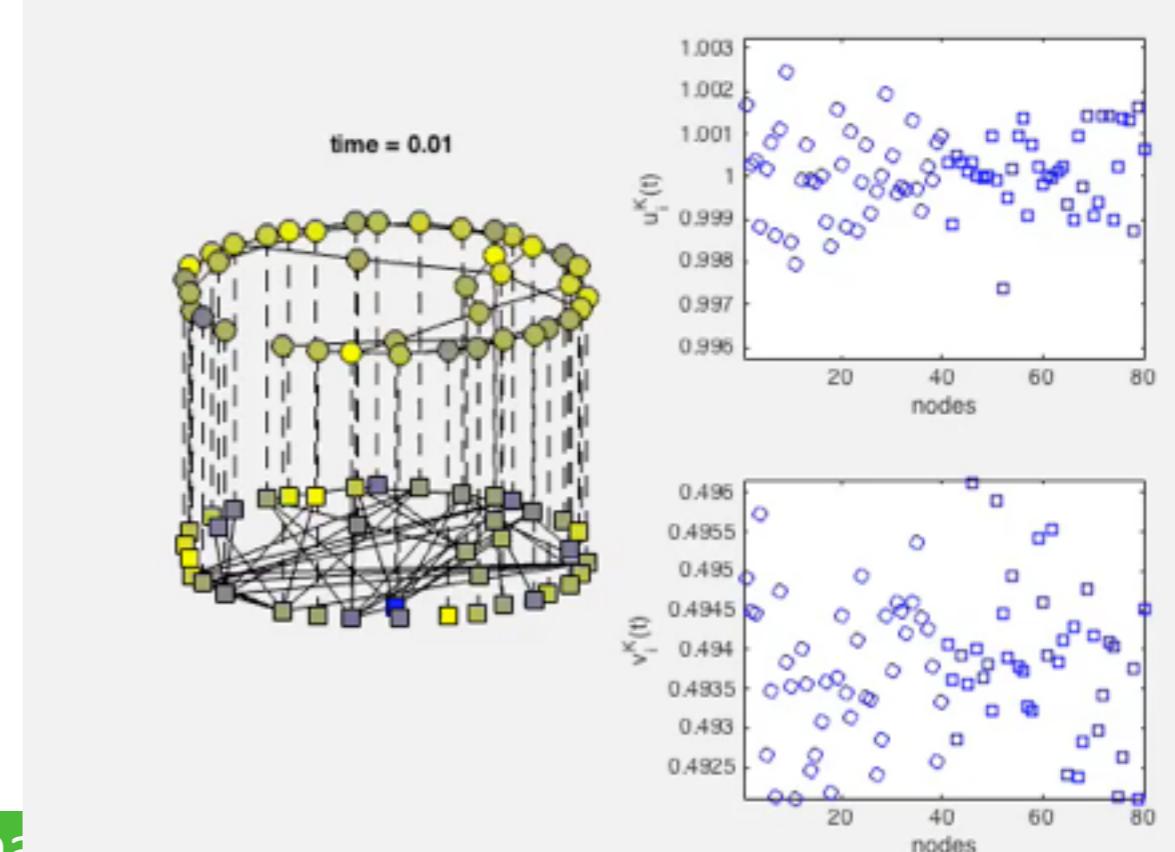
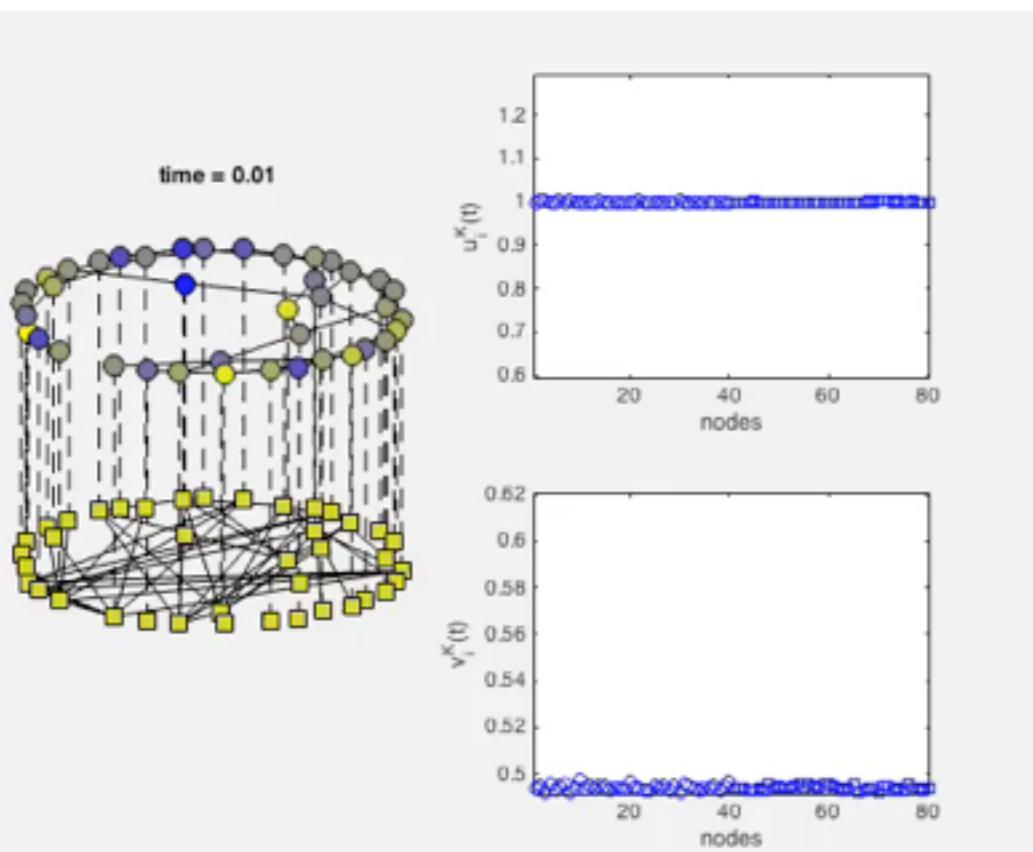
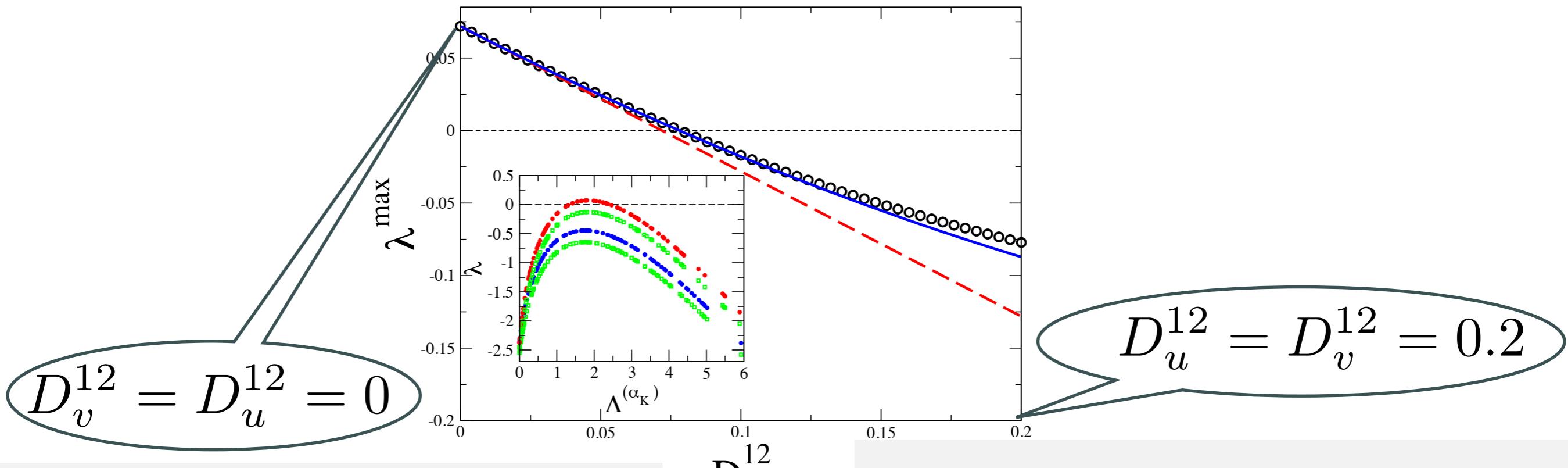
$$D_v^{12} = D_u^{12} = 0$$



$$D_u^{12} = 0 \quad D_v^{12} = 0.5$$

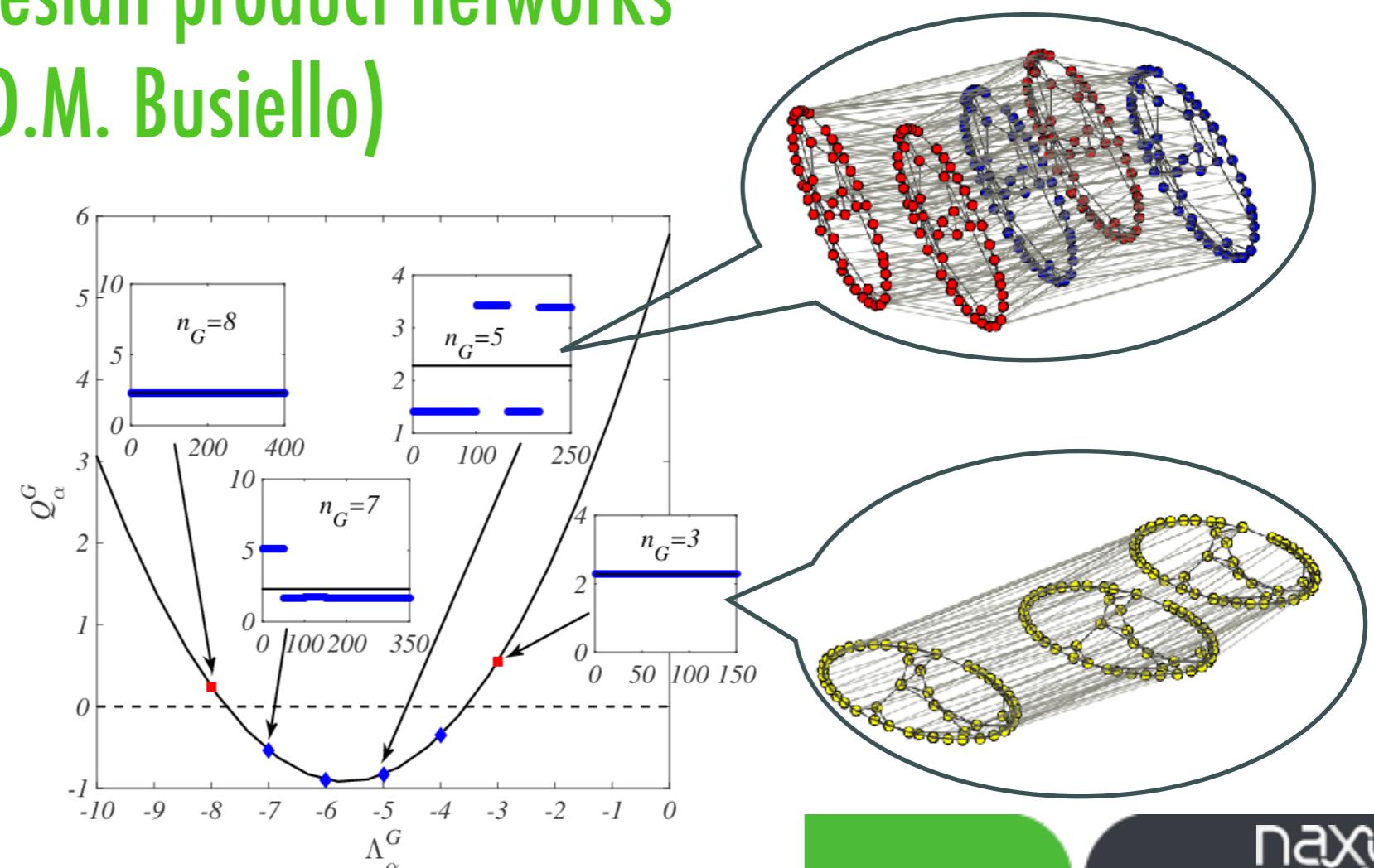


Small intra-layer diffusion case: destruction of patterns



Remarks

- 1) The complementary case - large intra-layer diffusion - can be handled as well
- 2) The completely degenerate case - same network on all layers - can be handled using Cartesian product networks
(see talk Friday morning D.M. Busiello)



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