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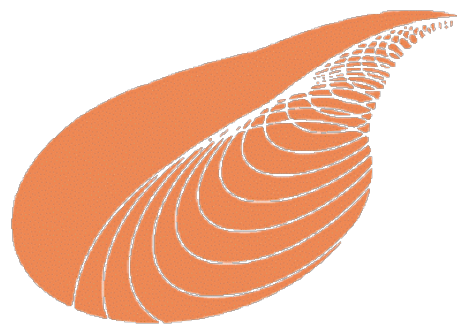
A journey in the zoo of Turing patterns



Acknowledgements

“Belgian” team: M. Asllani, J. Petit, G. Planchon

“Italian” team: D. Fanelli, D.M. Busiello, C. Cianci, M. Galanti, F. Di Patti



IAP VII/19 - DYSCO



Pattern ? [ref. Oxford dictionary]

pattern 

★ Top 1000 frequently used words

Pronunciation: /'pat(ə)n/ 

NOUN

- 1 A repeated decorative design:
'a neat blue herringbone pattern'

– More example sentences

'Included are geometrics, florals and foliates, animals and nature motifs and other decorative repeat patterns.'

'These aspects then become ornamented with Islamic-inspired decorative patterns and Islamic cultural artifacts.'

'It featured exuberant decorative patterns, designs in the brickwork and wooden attachments.'

- 1 A repeated decorative design:
'a neat blue herringbone pattern'

+ More example sentences

+ Synonyms

- 1.1 An arrangement or design regularly found in comparable objects:
'the house had been built on the usual pattern'

– More example sentences

'Structurally, the tumor cells were arranged in a medullary pattern composed of polygonal tumor cells.'

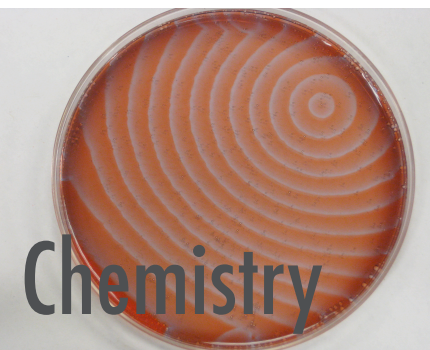
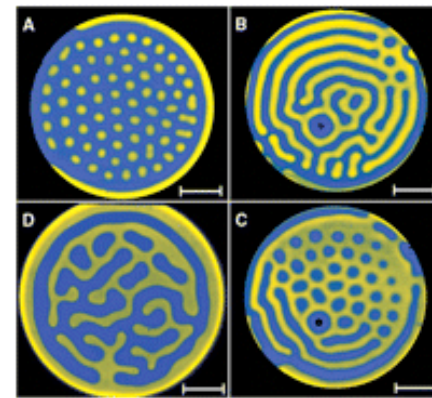
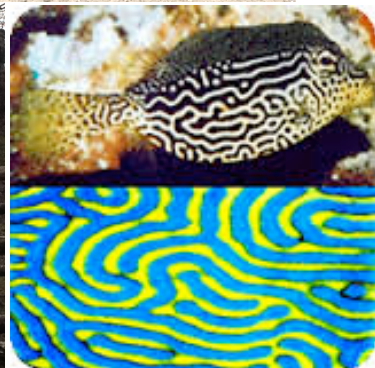
'The hair-cells within the spiralling cochlear duct are arranged in a pattern like the bristles of a brush.'

'The fossils indicate the wings had feathers, arranged in a similar pattern to that of modern birds.'

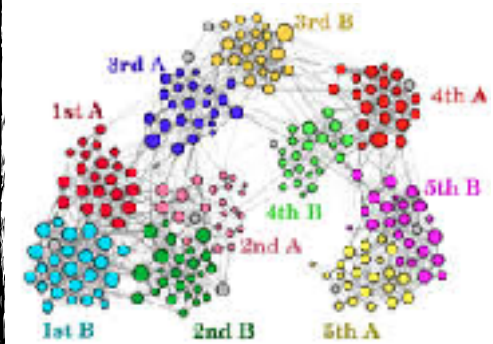
Patterns are ubiquitous



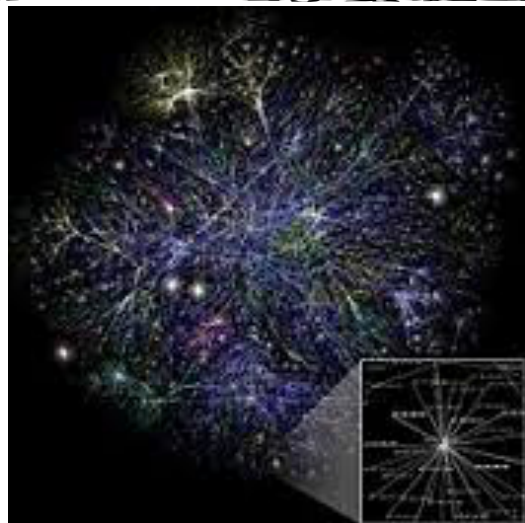
Animal kingdom



Chemistry



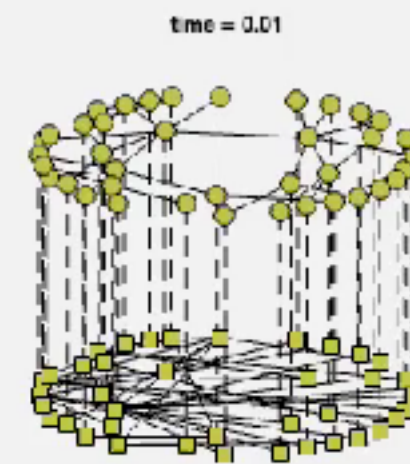
SocioPatterns



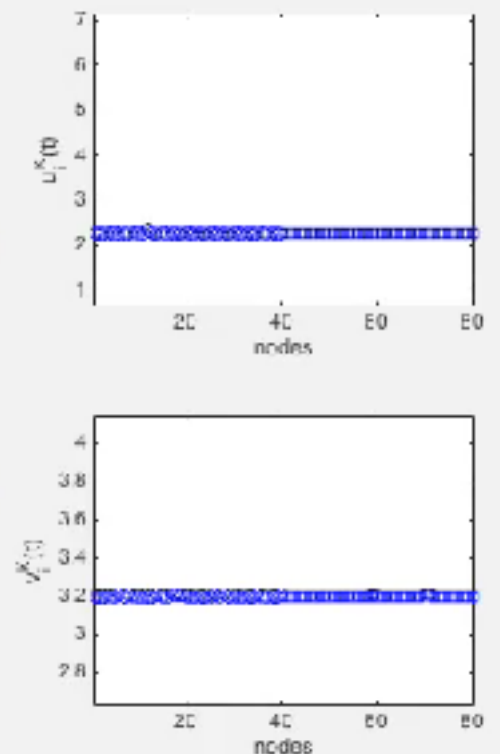
Internet



Twitter



Math models



One possible mechanism: Turing instability

Reaction

Autocatalysis
(positive feedback)

Inhibition
(negative feedback)

Activator

Inhibitor

Catalysis
(positive feedback)



Diffusion

$u(x, y, t)$: 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

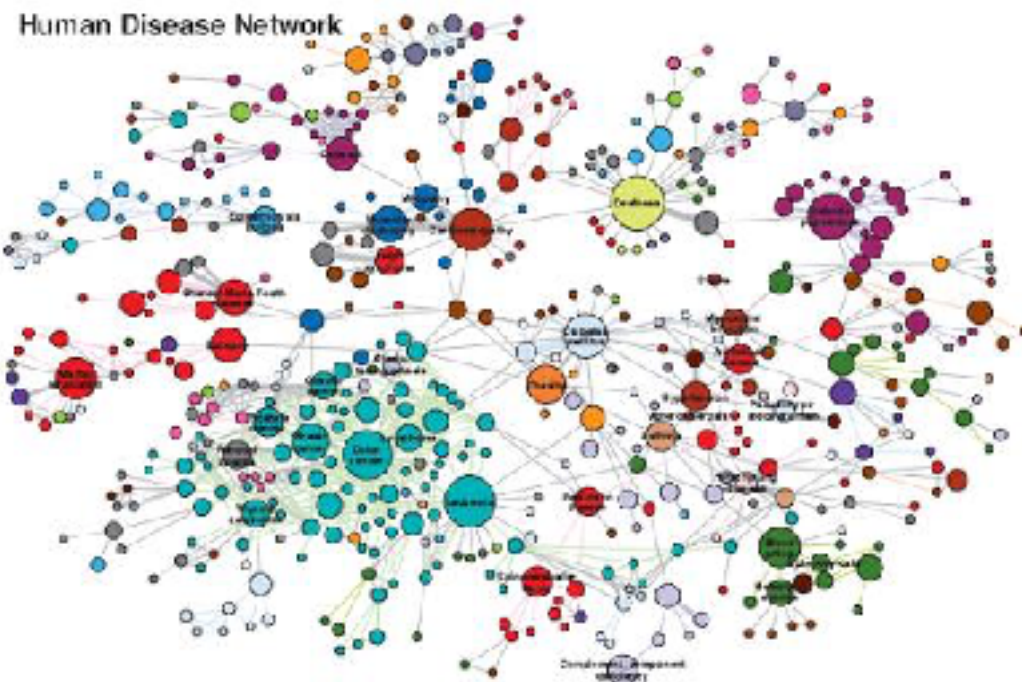
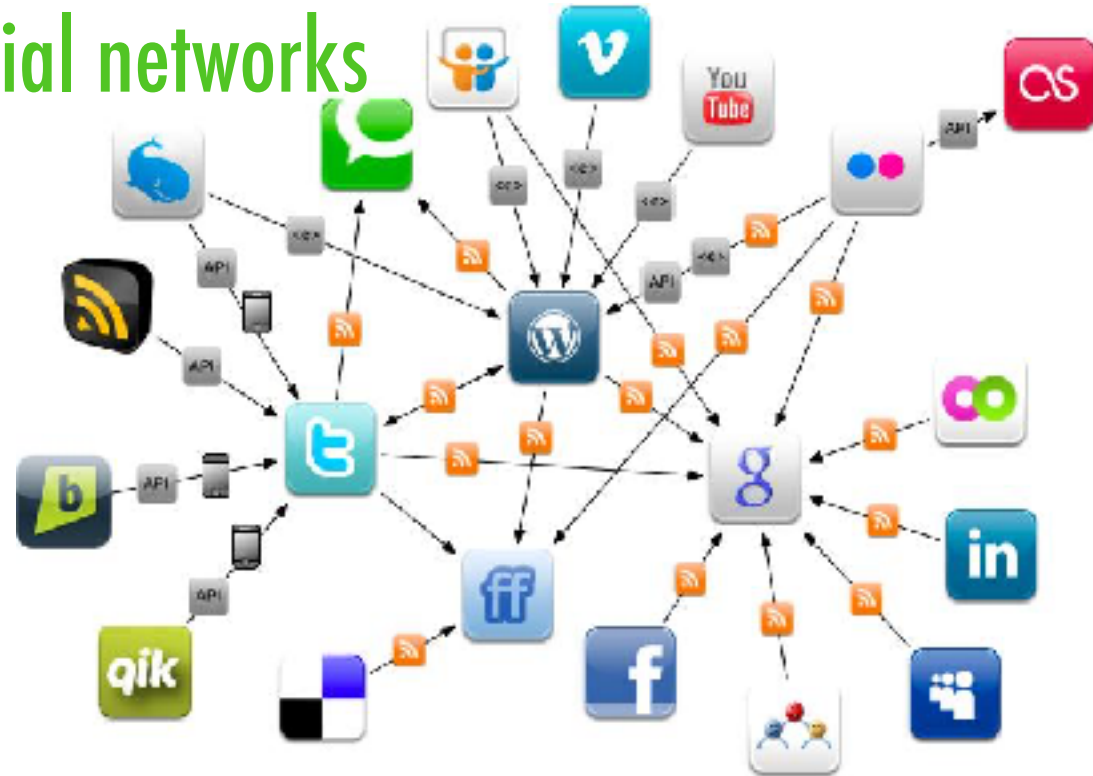
A.M.Turing, *The chemical basis of morphogenesis*, Phil. Trans. R Soc London B, **237**, (1952), pp.37

Networks are everywhere

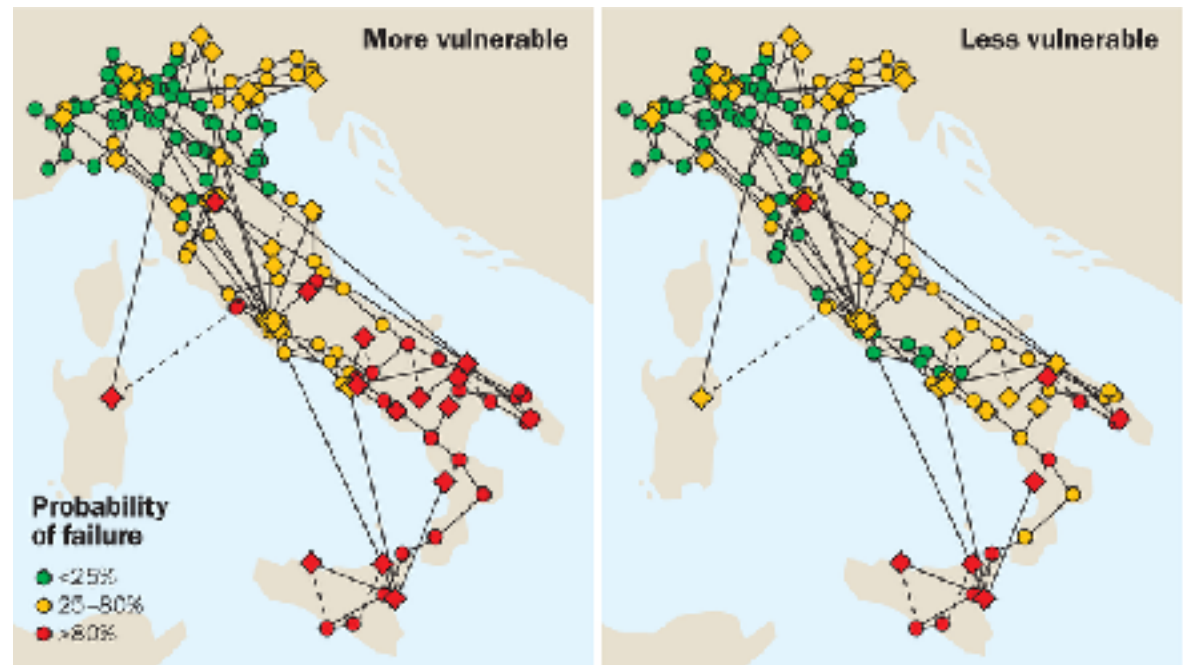


world flights map

social networks



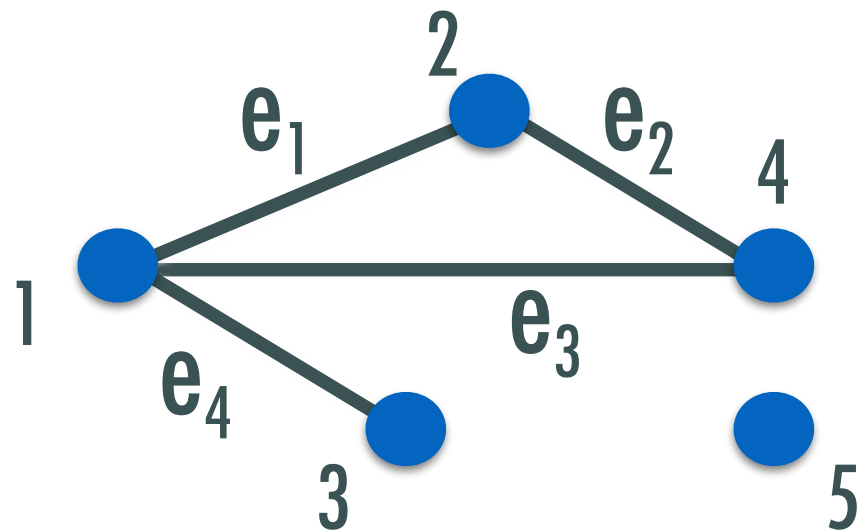
proteins networks



technological networks

(complex) Networks: some definitions

A network is a set of nodes connected by links (edges)



Ex.: 5 nodes and 4 edges (undirected)

Adjacency matrix

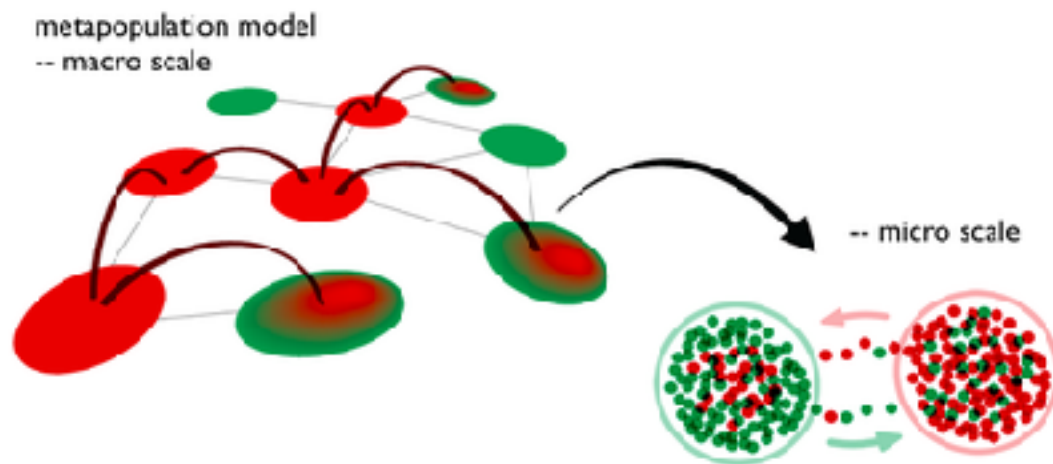
$$A_{ij} = \begin{cases} 1 & \text{if nodes } i \text{ and } j \text{ are linked} \\ 0 & \text{otherwise} \end{cases}$$

The number of links entering (going out) from each node is called in-degree (out-degree)

Ex.: degree node 1 = 3
degree nodes 2 & 4 = 2
degree node 3 = 1
degree node 5 = 0

A network is said to be complex if the degree distribution is not trivial, i.e. not constant (lattice) nor Poissonian (random, Erdős-Rényi)

Extension to networks



Metapopulation models

e.g. in the framework of ecology:

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

Reactions occur at each node. Diffusion occurs across edges.

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

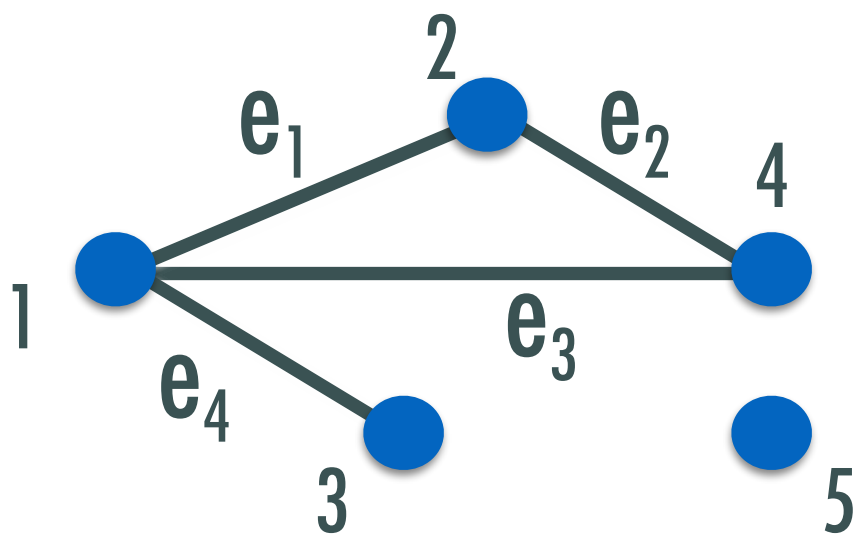
Reaction term:

$$\begin{cases} \dot{u}_i(t) &= f(u_i(t), v_i(t)) \\ \dot{v}_i(t) &= g(u_i(t), v_i(t)) \end{cases} \quad \forall i = 1, \dots, n \text{ and } t > 0.$$

At each node $i=1,\dots,n$, “species” u and v react through some non-linear functions f and g depending on the quantities available at node i -th (metapopulation assumption)

Diffusion term:

Diffusive transport of species into a certain node i is given by the sum of incoming fluxes to node i from other connected nodes j , fluxes are proportional to the concentration difference between the nodes (Fick's law).



Ex.: consider the amount of u in node 1,
 u can enter from 2, 3 and 4
 u can leave 1 to go to 2, 3 and 4

$$u_2 + u_3 + u_4 - 3u_1 = \sum_j A_{1j}u_j - k_1u_1 = \sum_j (A_{1j} - \delta_{1j}k_j)u_j := \sum_j L_{1j}u_j$$

L is called Laplacian matrix of the network

The model:

$$\begin{cases} \dot{u}_i(t) &= f(u_i(t), v_i(t)) + D_u \sum_{j=1}^n L_{ij} u_j(t) \\ \dot{v}_i(t) &= g(u_i(t), v_i(t)) + D_v \sum_{j=1}^n L_{ij} v_j(t) \end{cases} \quad \forall i = 1, \dots, n \text{ and } t > 0.$$

D_u and D_v are the diffusion coefficients of species u and v

Observe that because the network is undirected, the matrices A and L are symmetric

General strategy for the network case

1) Assume there exists a spatially homogeneous solution:

$$(u_i, v_i) = (\hat{u}, \hat{v}) \quad \forall i = 1, \dots, n$$

which moreover is stable when there is no diffusion:

$$D_u = D_v = 0$$

2) Linearize around this solution

$$\begin{aligned} u_i &= \hat{u} + \delta u_i \\ v_i &= \hat{v} + \delta v_i \end{aligned} \quad \begin{pmatrix} \dot{\delta u} \\ \dot{\delta v} \end{pmatrix} = \tilde{\mathcal{J}} \begin{pmatrix} \delta u \\ \delta v \end{pmatrix}$$

$$\tilde{\mathcal{J}} = \begin{pmatrix} f_u \mathbf{I}_n + D_u \mathbf{L} & f_v \mathbf{I}_n \\ g_u \mathbf{I}_n & g_v \mathbf{I}_n + D_v \mathbf{L} \end{pmatrix}$$

General strategy for the network case

3) Prove that (possibly) the spatially homogeneous solution:

$$(u_i, v_i) = (\hat{u}, \hat{v}) \quad \forall i = 1, \dots, n$$

turns out to be unstable once the diffusion is in action

$$D_u > 0 \text{ and } D_v > 0$$

Sketch of the proof

i) Let $L\vec{\phi}^\alpha = \Lambda^\alpha \vec{\phi}^\alpha$, $\alpha = 1, \dots, n$ $\vec{\phi}^\alpha = (\phi_1^\alpha, \dots, \phi_n^\alpha)$

$$\sum_i \phi_i^\alpha \phi_i^\beta = \delta_{\alpha\beta} \quad \Lambda^\alpha \leq 0$$

ii) decompose the solution on the eigenbasis and use the ansatz

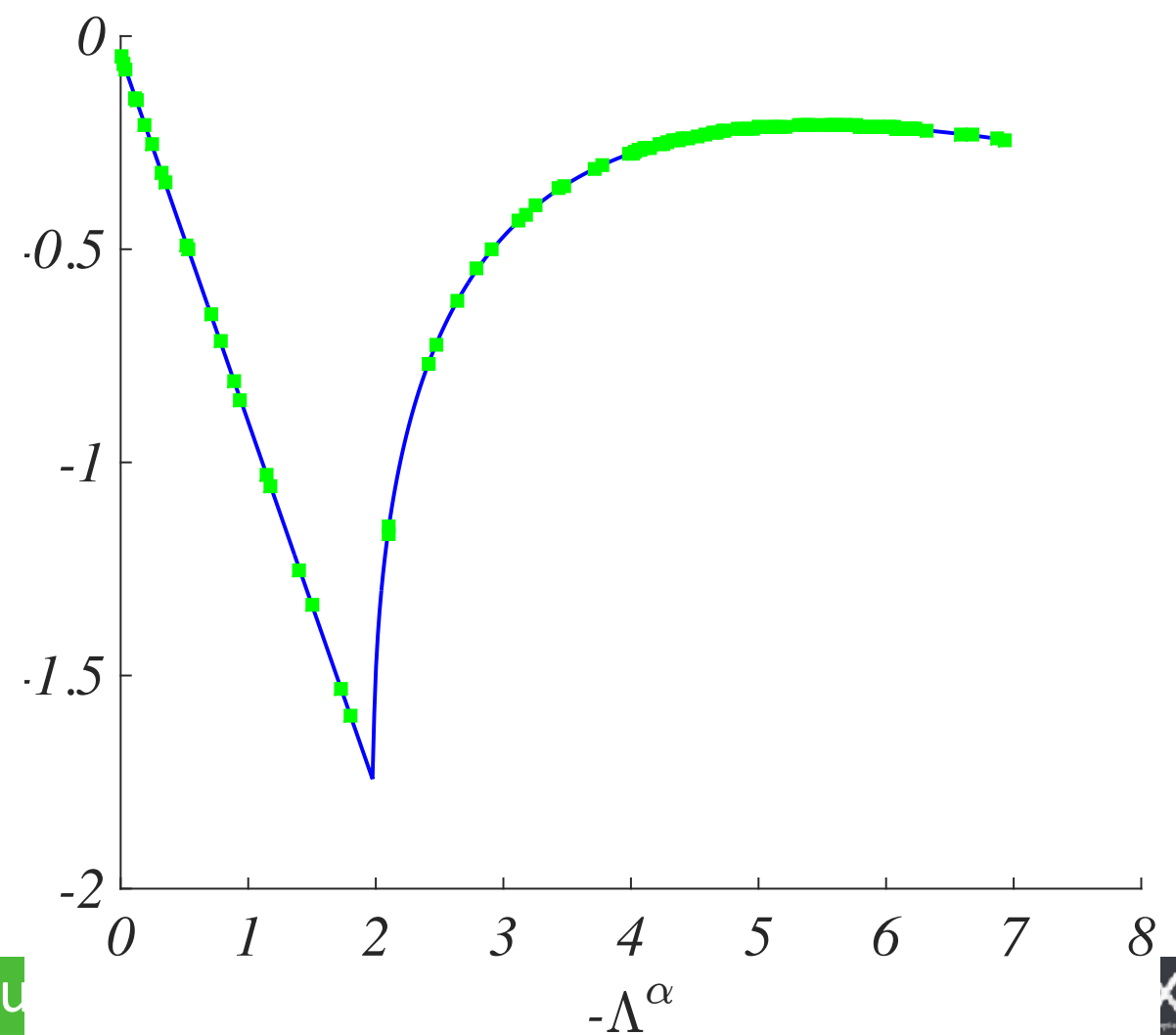
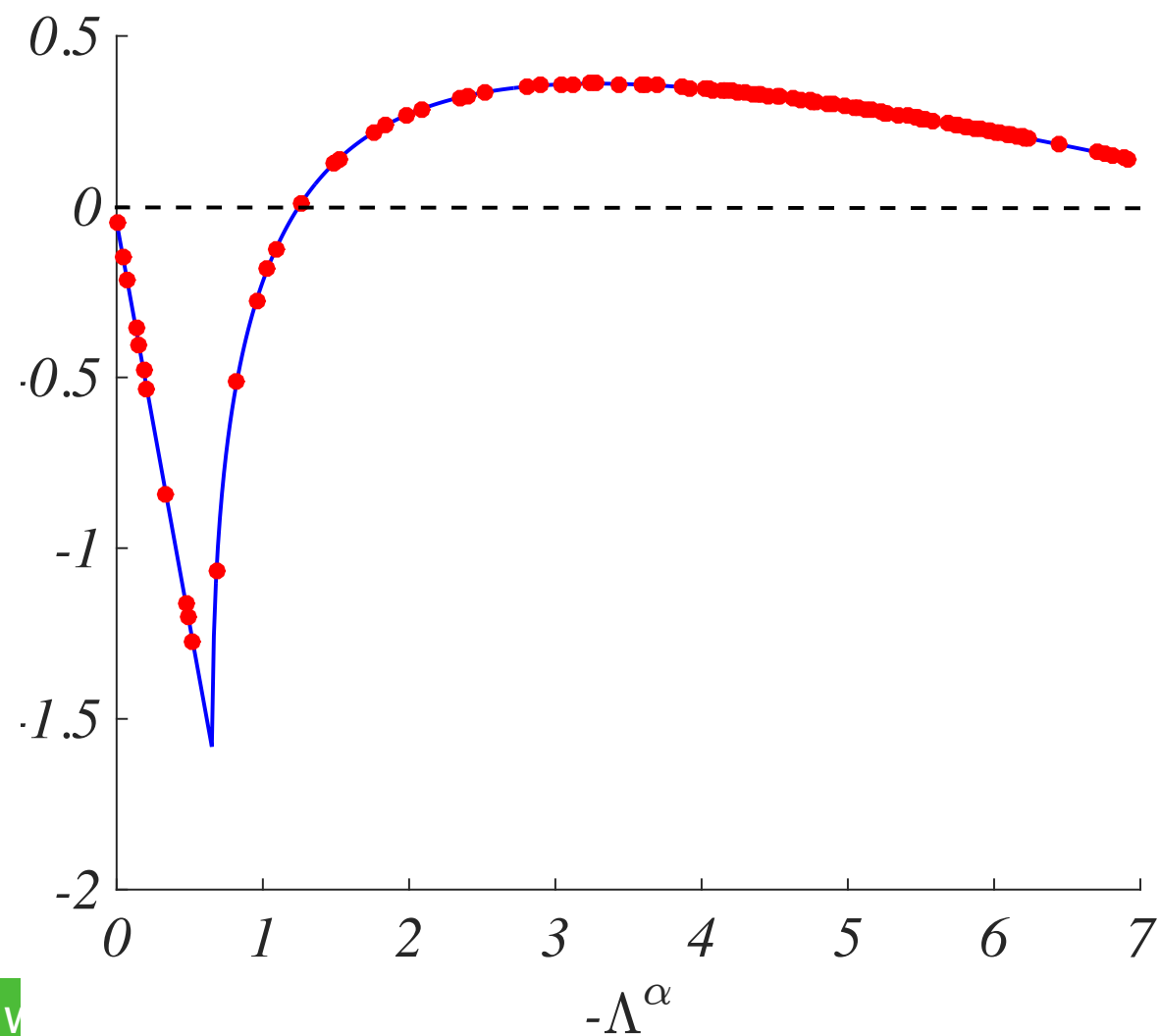
$$\delta u_i(t) = \sum_{\alpha=1}^n c_\alpha \phi_i^\alpha e^{\lambda_\alpha t}$$

General strategy

iii) λ_α (called relation dispersion) is solution of

$$\det \left[\lambda_\alpha - \begin{pmatrix} f_u + D_u \Lambda^\alpha & f_v \\ g_u & g_v + D_v \Lambda^\alpha \end{pmatrix} \right] = 0$$

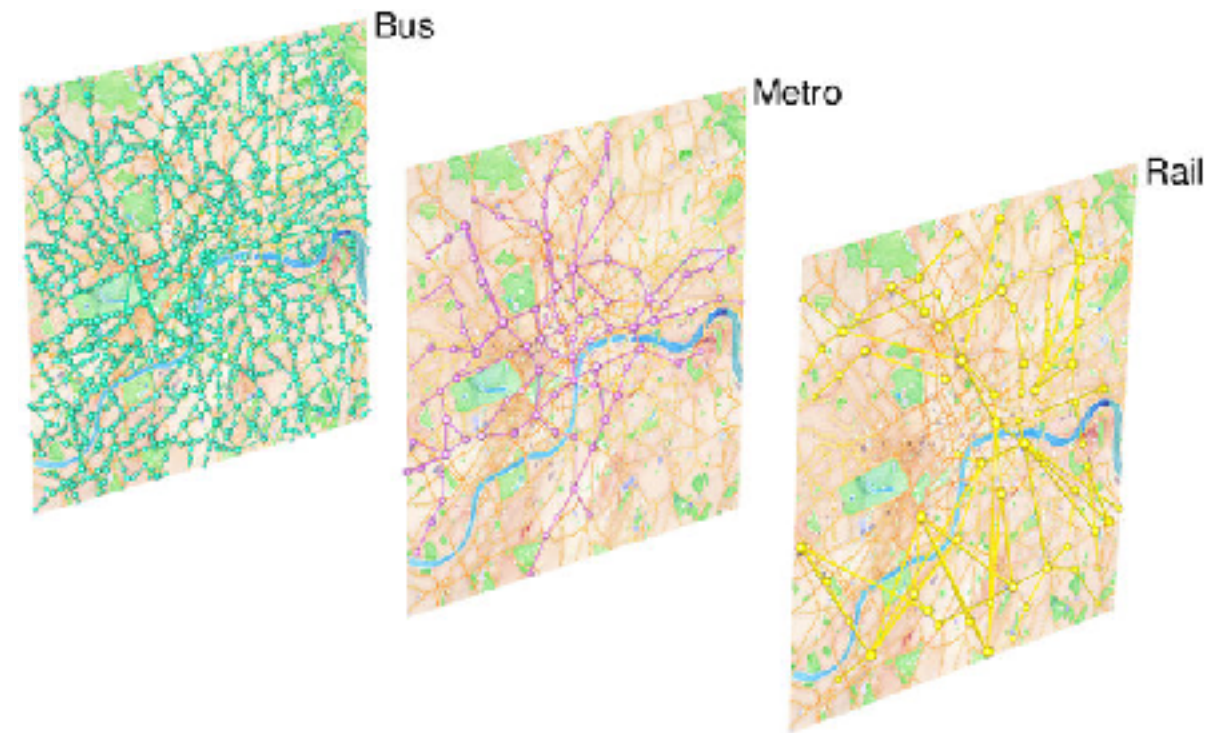
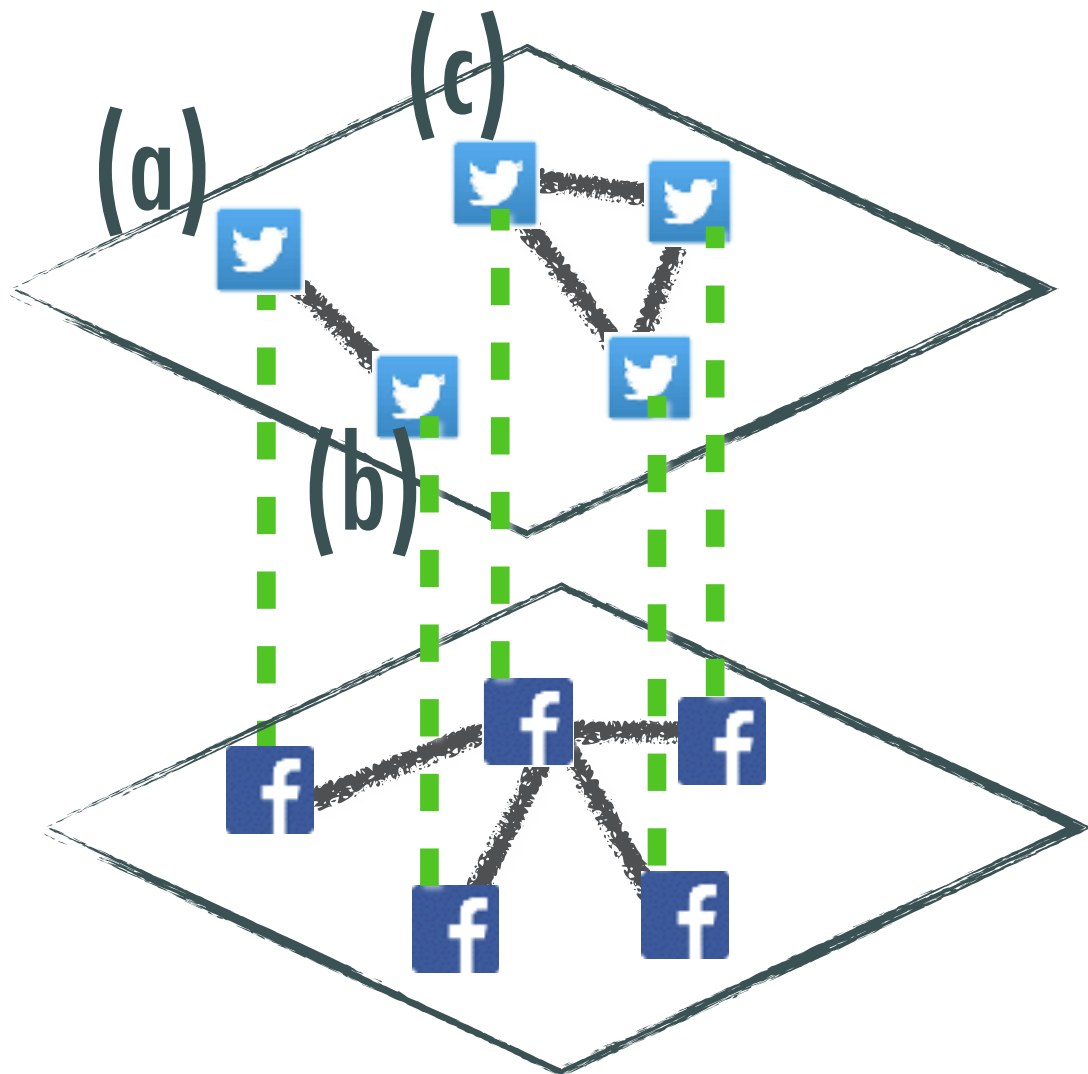
iv) if there exists Λ^{α_c} such that $\Re \lambda_{\alpha_c} > 0$ then Turing patterns do emerge.



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

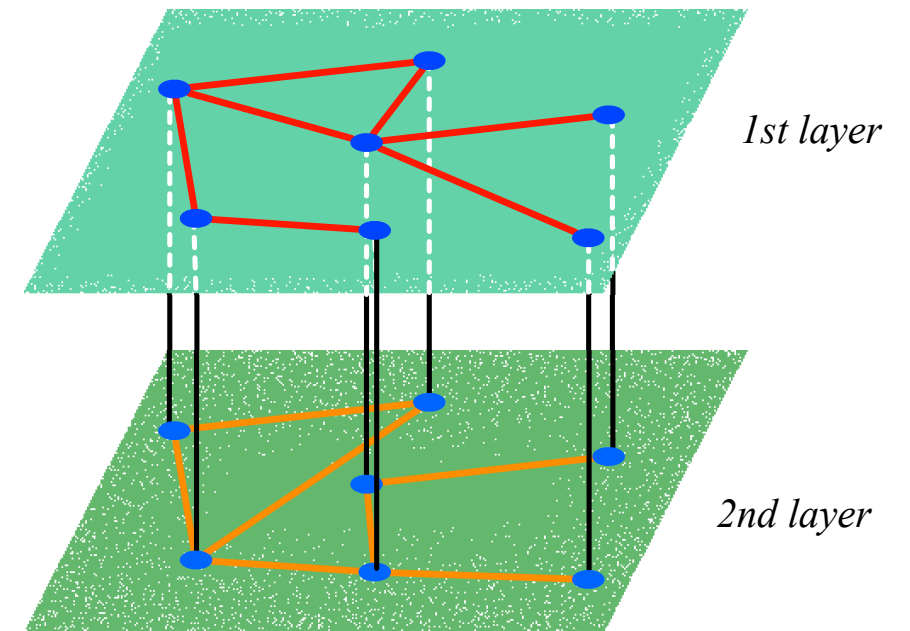
Turing instabilities on multiplex networks

adjacency matrix of
layer K

degree of i-th node
in layer K

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

Laplacian matrix of
layer K



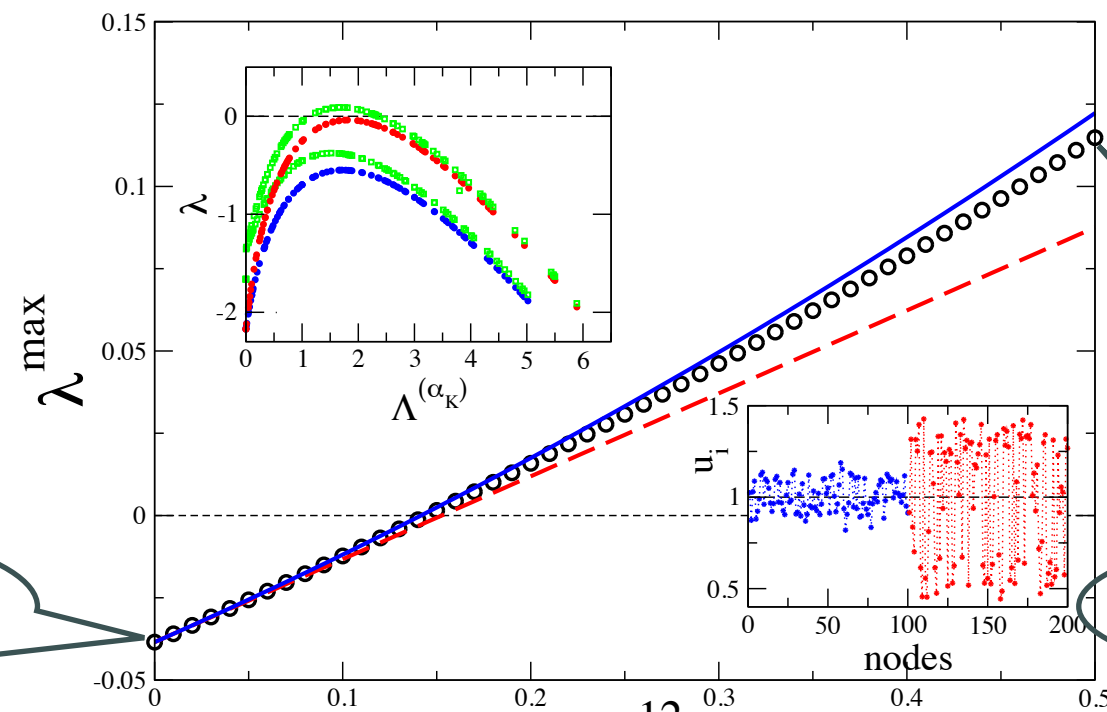
The same Ω nodes are present in each layer

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

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

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

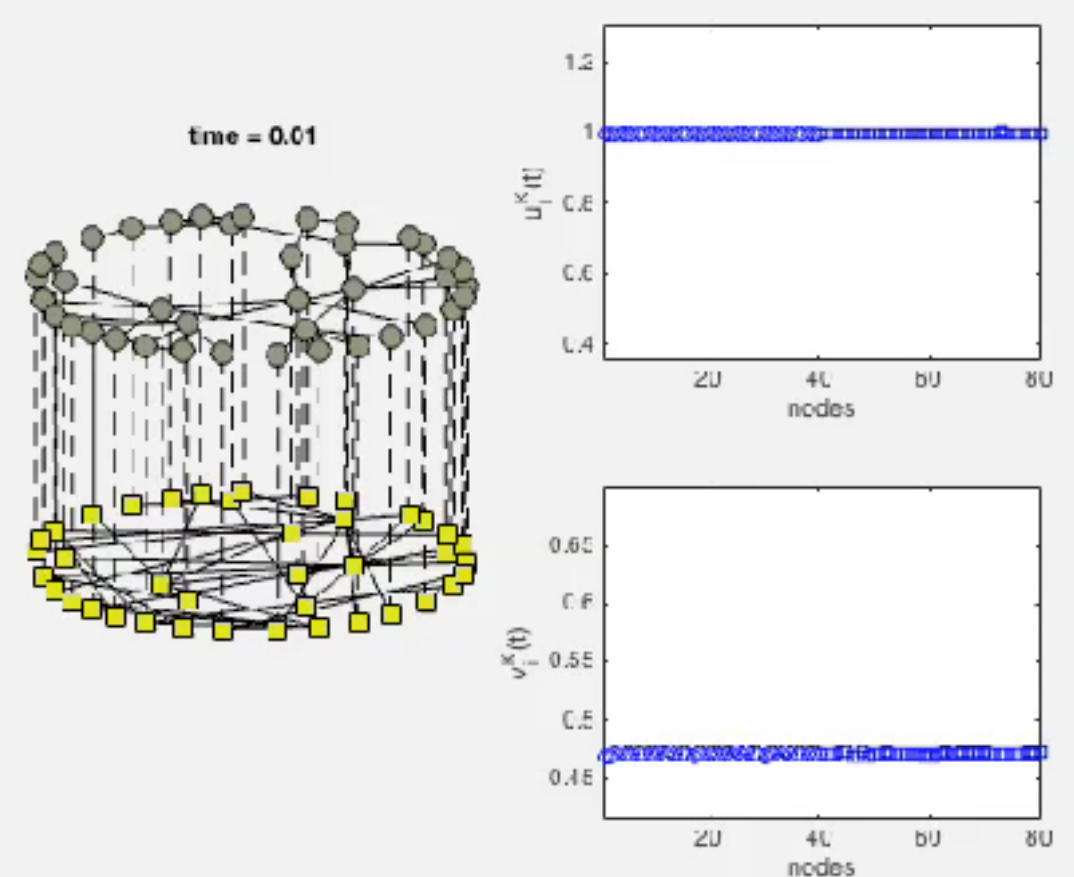
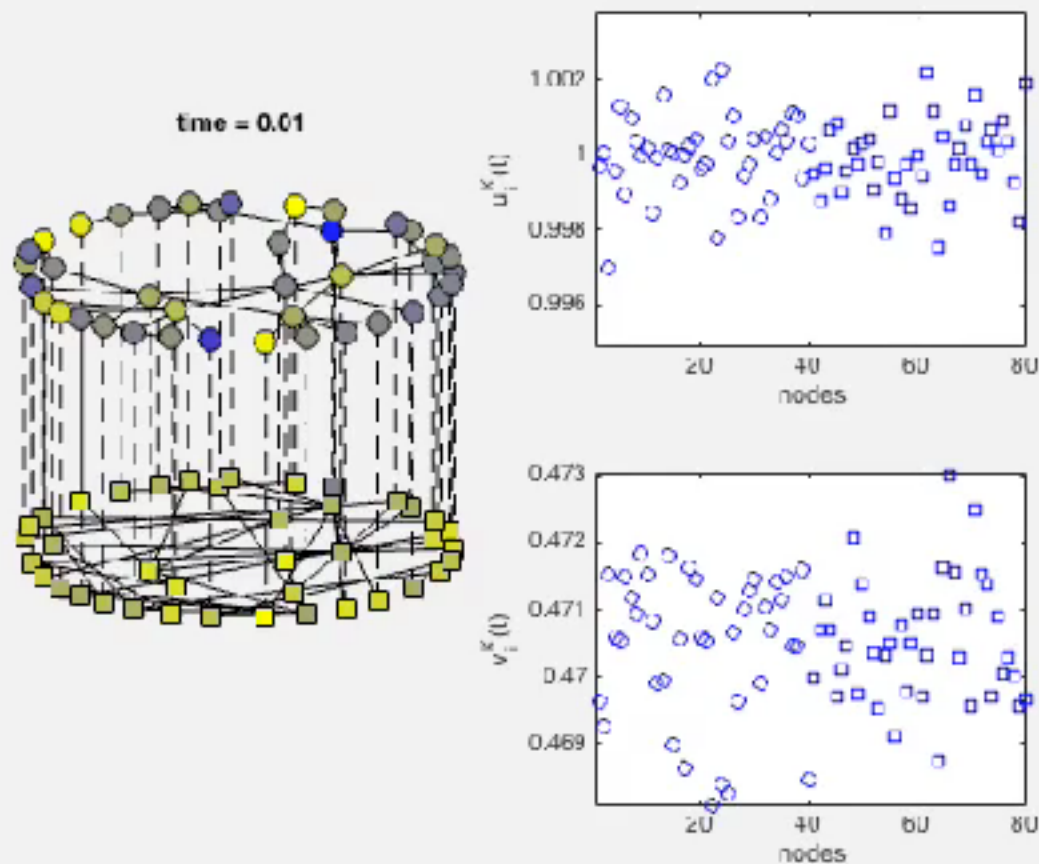
Small intra-layer diffusion case: onset of patterns



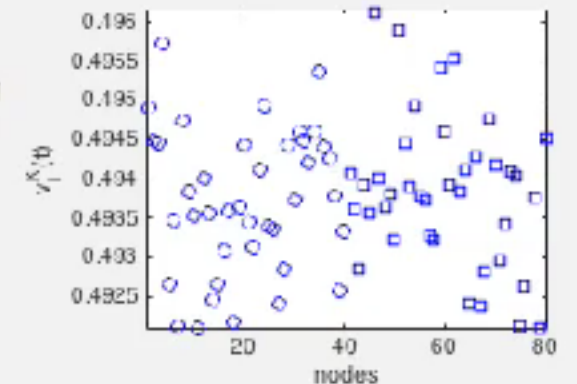
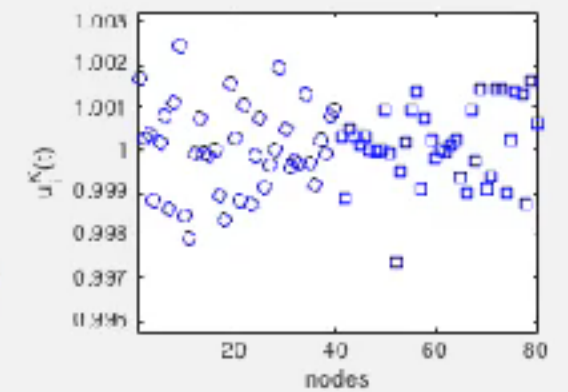
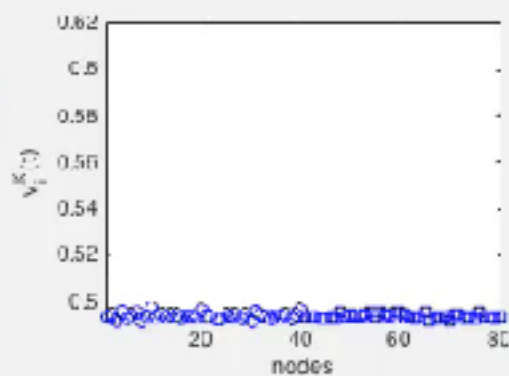
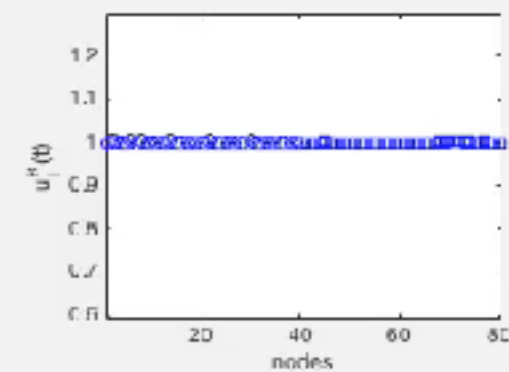
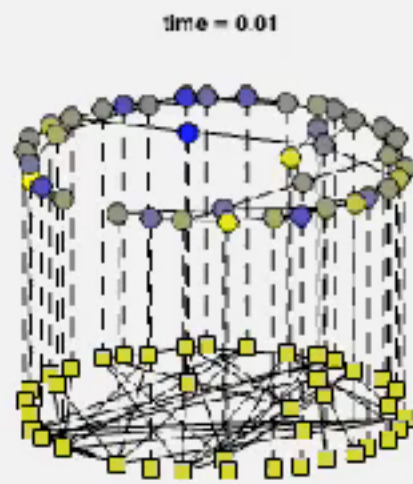
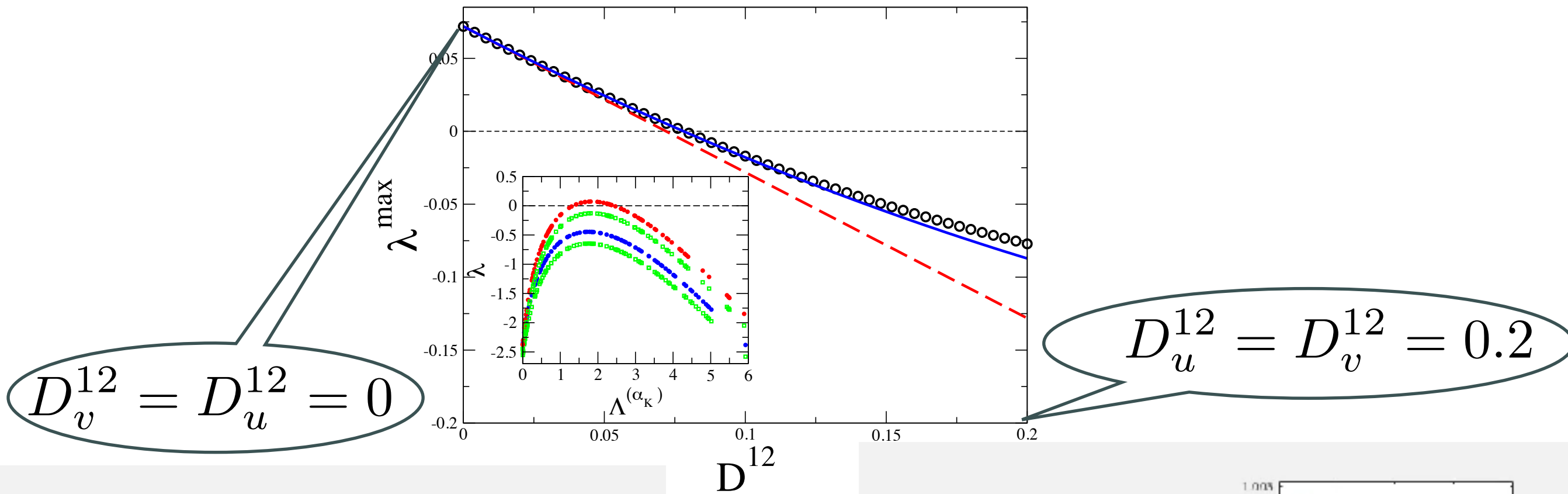
$$D_v^{12} = D_u^{12} = 0$$

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

D_v^{12}

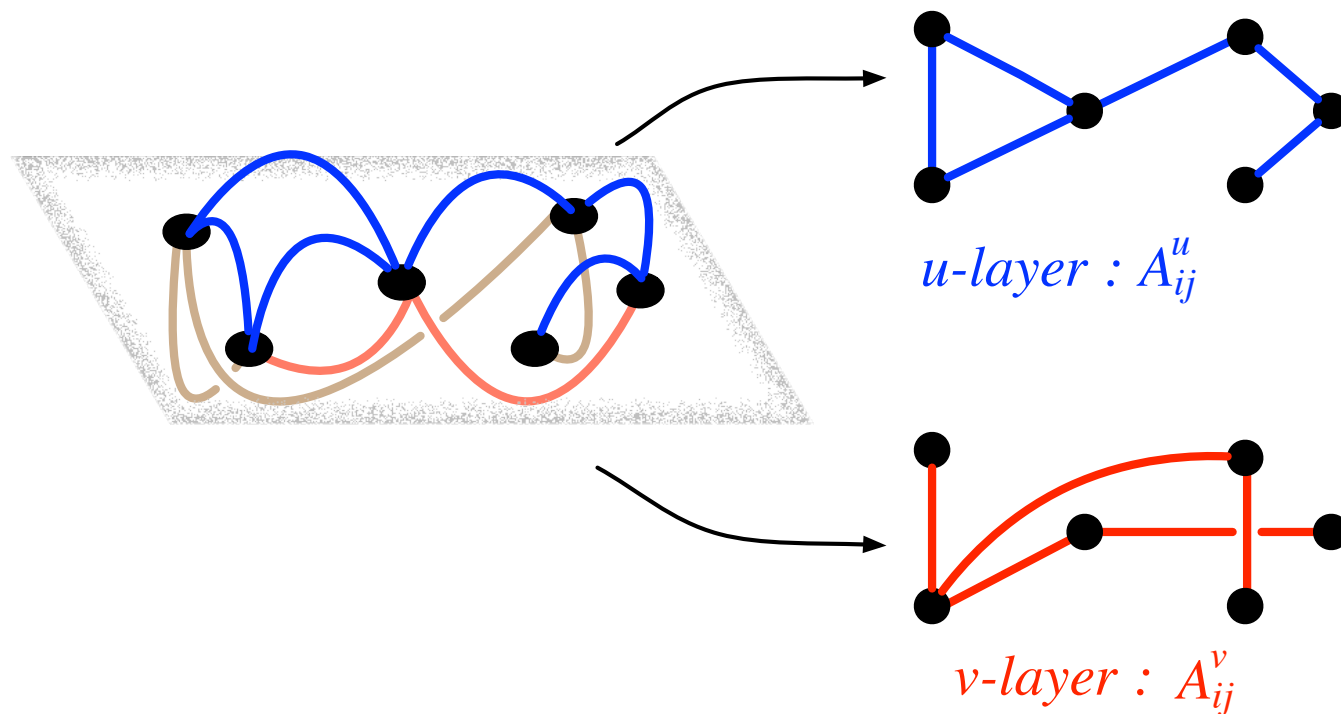


Small intra-layer diffusion case: destruction of patterns



Can we control the topology to create (destroy) patterns?

Let us consider a multigraph, e.g. two nodes can be connected through different edges



$$\epsilon = 0$$

$$A^u(0) = A^0$$

$$A^v(0) = A^0$$

$$A^u(\epsilon) = A^0 + \epsilon(A^1 - A^0)$$

$$A^v(\epsilon) = A^0 + \epsilon(A^2 - A^0)$$

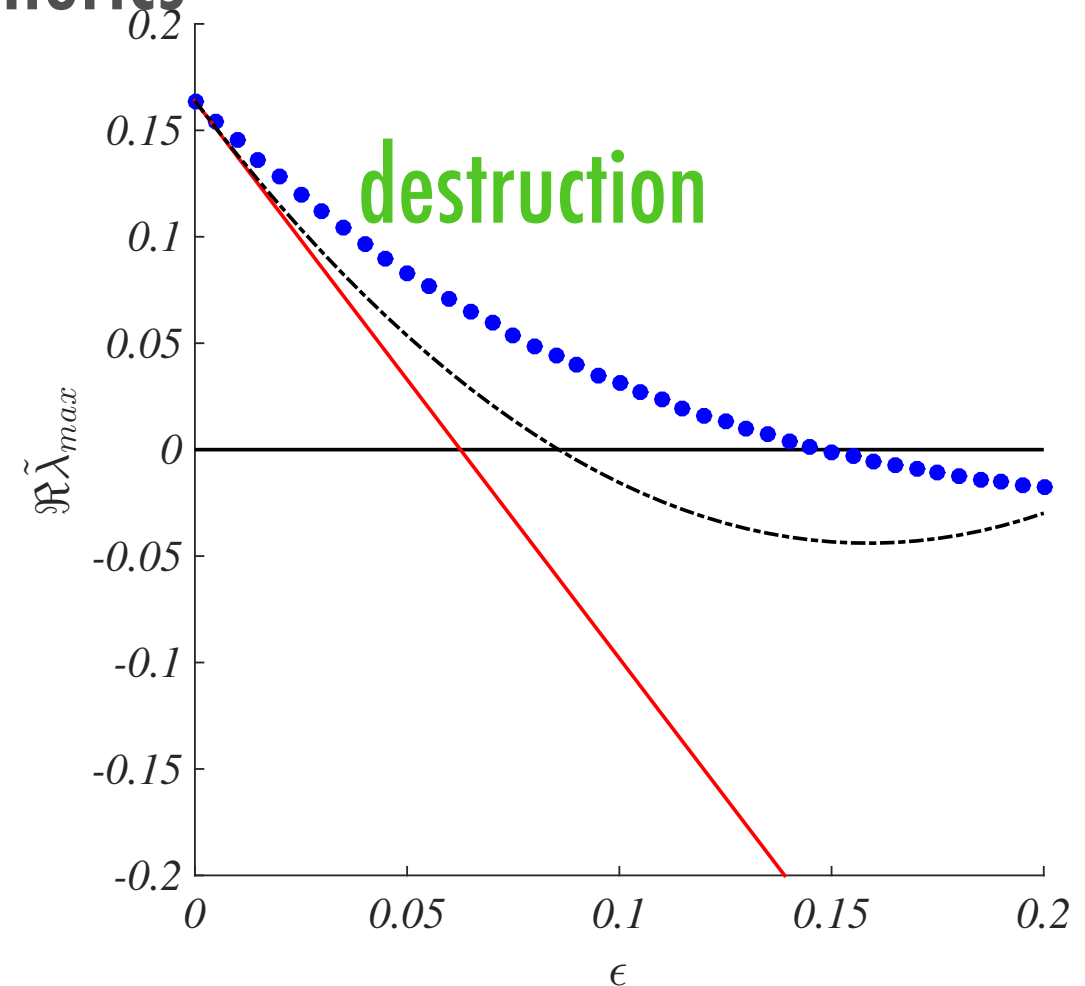
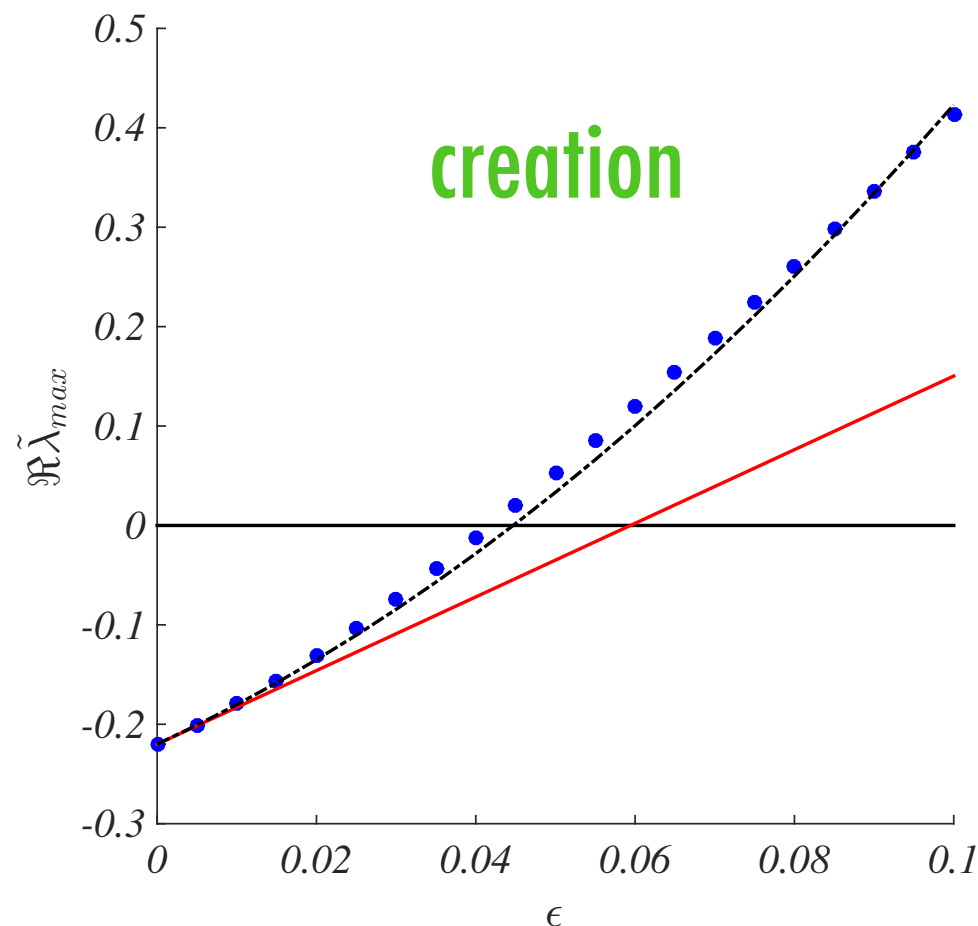
$$\epsilon = 1$$

$$A^u(1) = A^1$$

$$A^v(1) = A^2$$

Can we control the topology to create (destroy) patterns?

theory vs numerics



$\epsilon = 0$

$\epsilon = 1$

$$A^u(0) = A^0$$

$$A^u(\epsilon) = A^0 + \epsilon(A^1 - A^0)$$

$$A^u(1) = A^1$$

$$A^v(0) = A^0$$

$$A^v(\epsilon) = A^0 + \epsilon(A^2 - A^0)$$

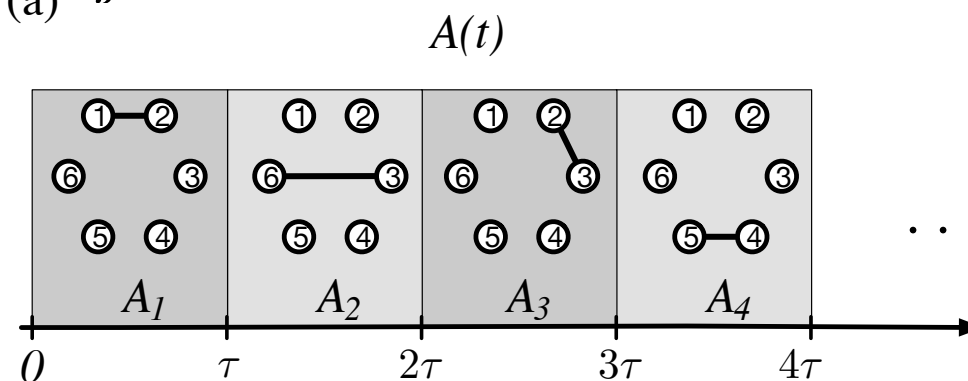
$$A^v(1) = A^2$$

Temporal networks

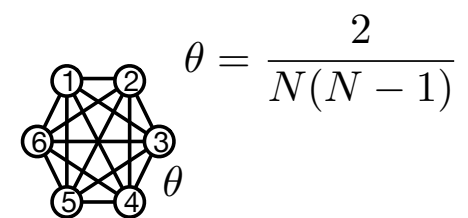
$$\dot{u}_i(t) = f(u_i, v_i) + D_u \sum_{j=1}^N L_{ij}(t/\epsilon) u_j(t)$$

$$\dot{v}_i(t) = g(u_i, v_i) + D_v \sum_{j=1}^N L_{ij}(t/\epsilon) v_j(t)$$

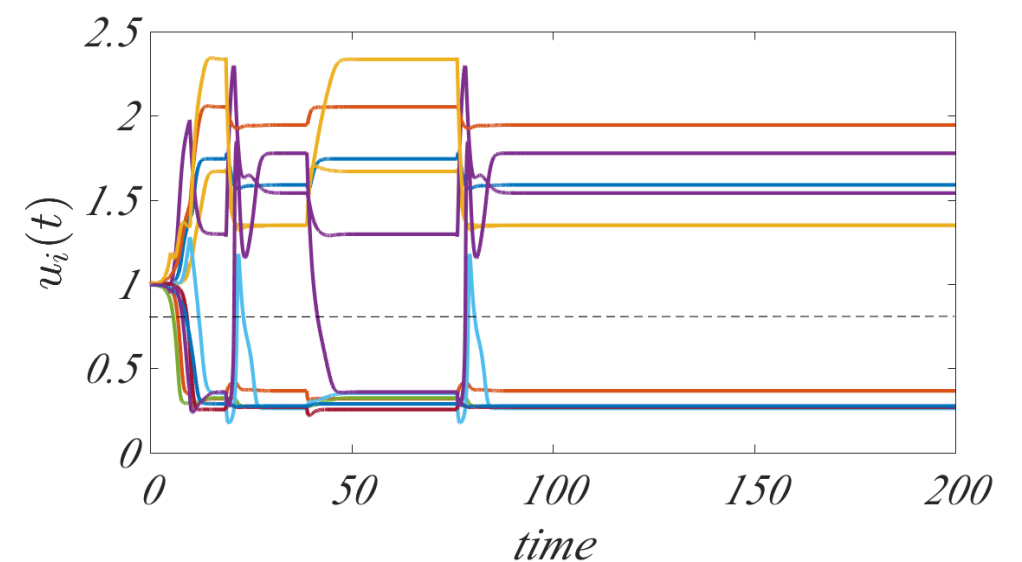
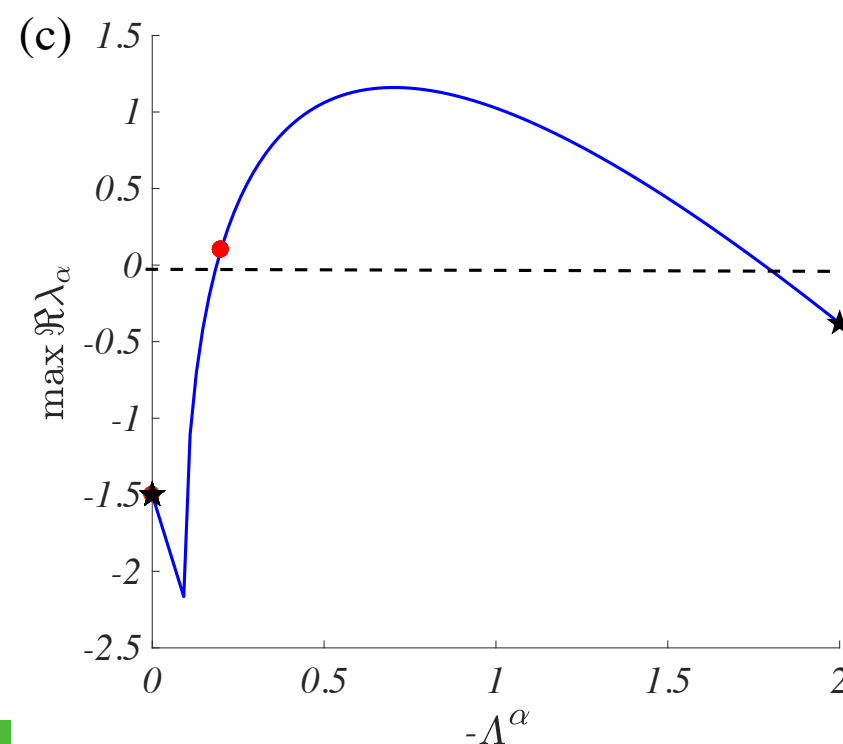
(a)



$\langle A \rangle$



(b)



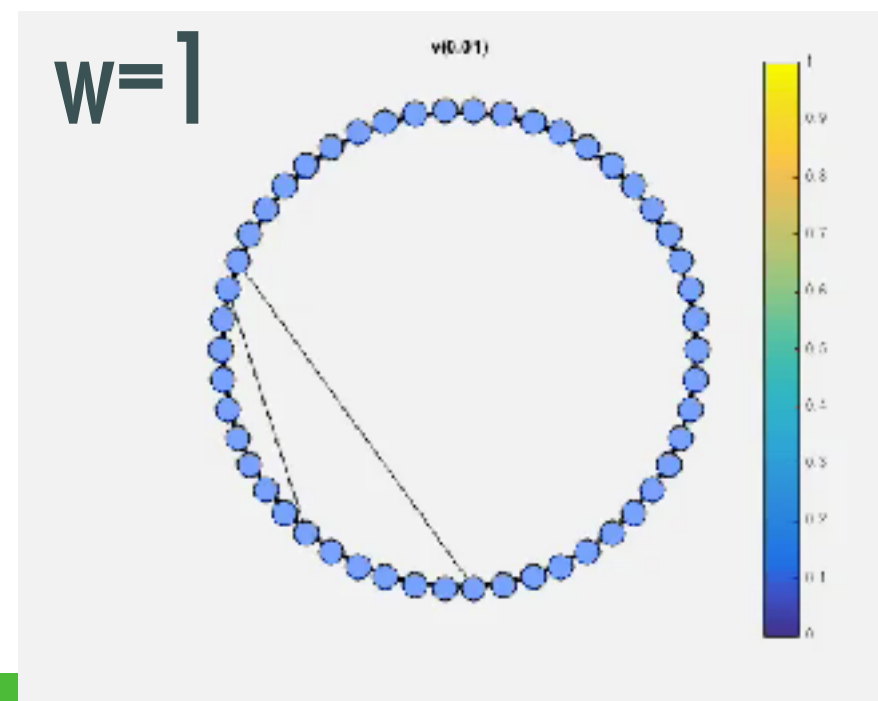
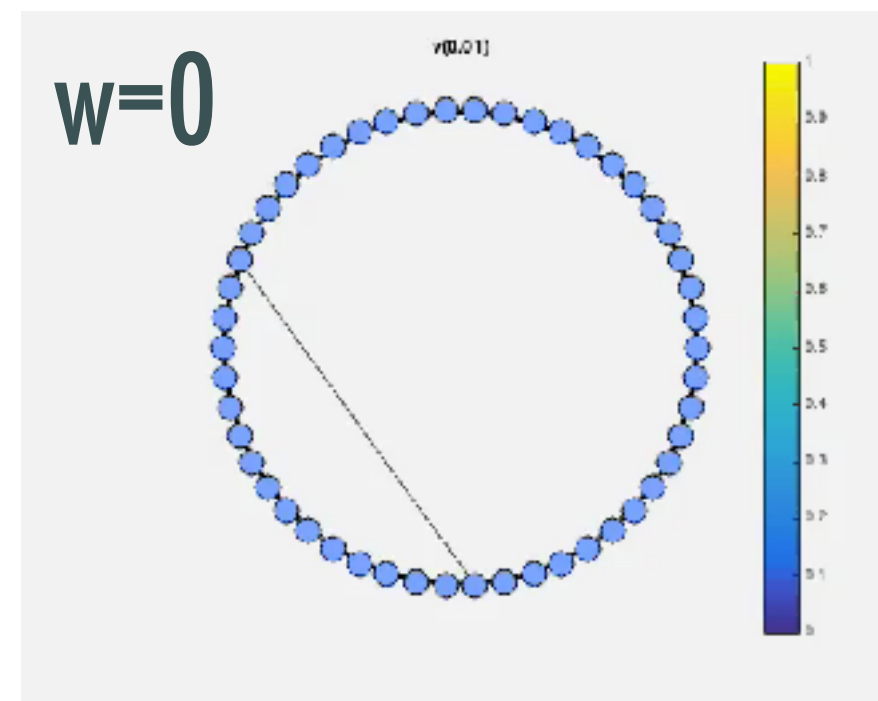
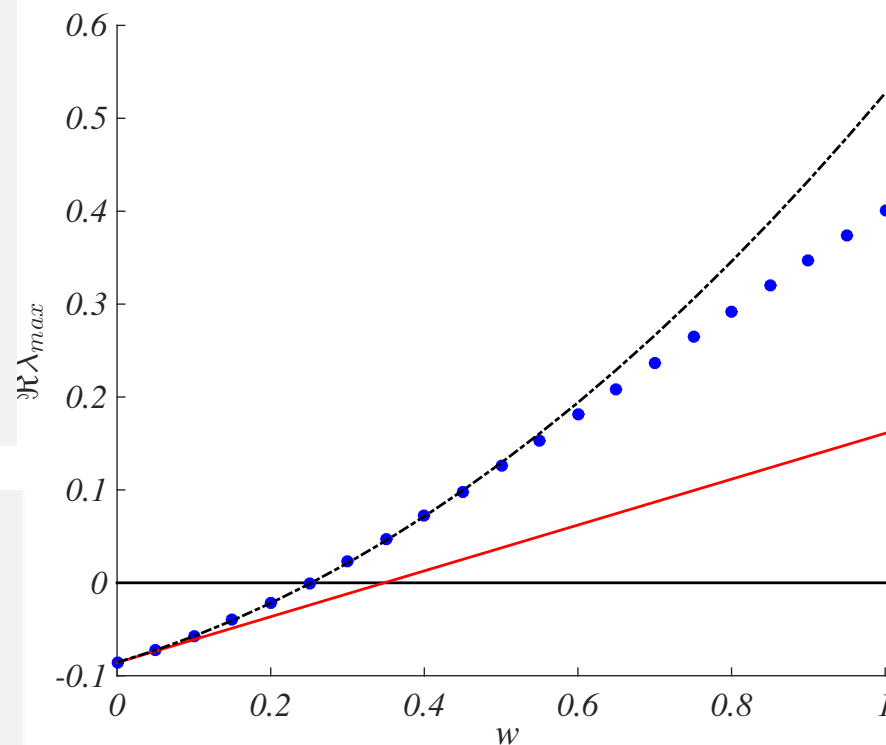
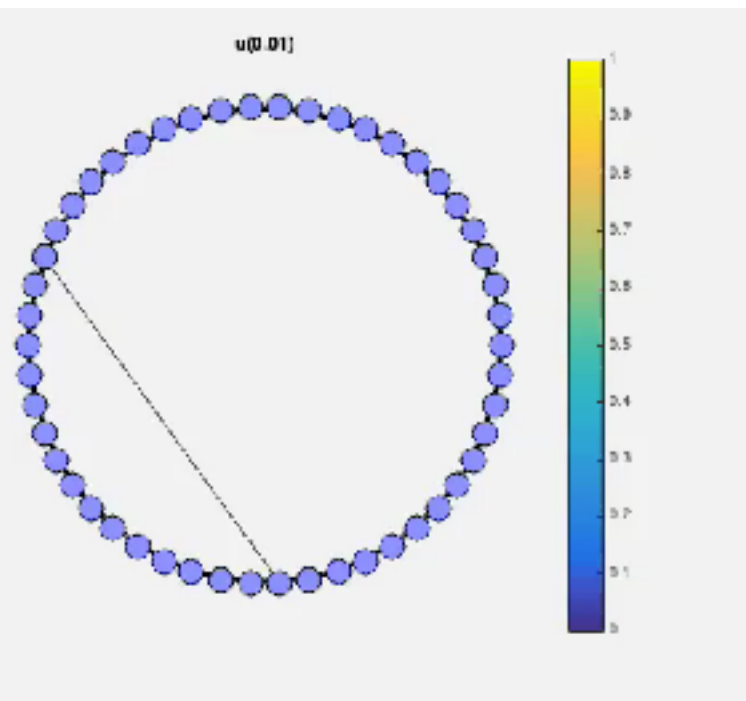
(d)

Can we control the topology to create (destroy) patterns?

Create patterns by adding a single (optimally chosen) link

$$A^u(w) = A^0$$

$$A^v(w) = A^0 + wT^{(ij)}$$



Some papers

Theory of Turing Patterns on Time Varying Networks, J. Petit, B. Lauwens, D. Fanelli, T. Carletti, Physical Review Letters, **119**, pp. 148301-1–5, (2017)

Tune the topology to create or destroy patterns, M. Asllani, T. Carletti, D. Fanelli, European Physical Journal B. **89**, pp. 260 (2016)

Pattern formation in a two-component reaction-diffusion system with delayed processes on a network, J. Petit, M. Asllani, D. Fanelli, B. Lauwens, T. Carletti, Physica A, **462**, pp.230, (2016)

Delay induced Turing-like waves for one species reaction–diffusion model on a network, J. Petit, T. Carletti, M. Asllani, D. Fanelli, Europhysics Letters. **111**, 5, pp. 58002, (2015)

Turing instabilities on Cartesian product networks, M. Asllani, D.M. Busiello, T. Carletti, D. Fanelli, G. Planchon, Scientific Reports. **5**, pp. 12927, (2015)

Turing patterns in multiplex networks, M. Asllani, D.M. Busiello, T. Carletti, D. Fanelli, G. Planchon, Physical Review E ,**90**, 4, pp. 042814, (2014)