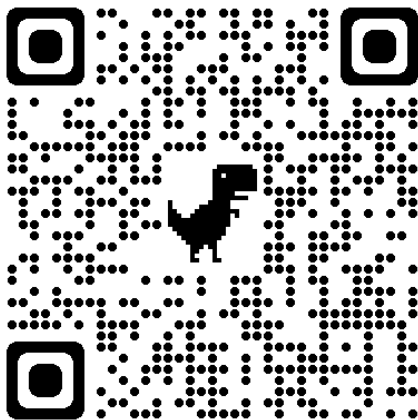
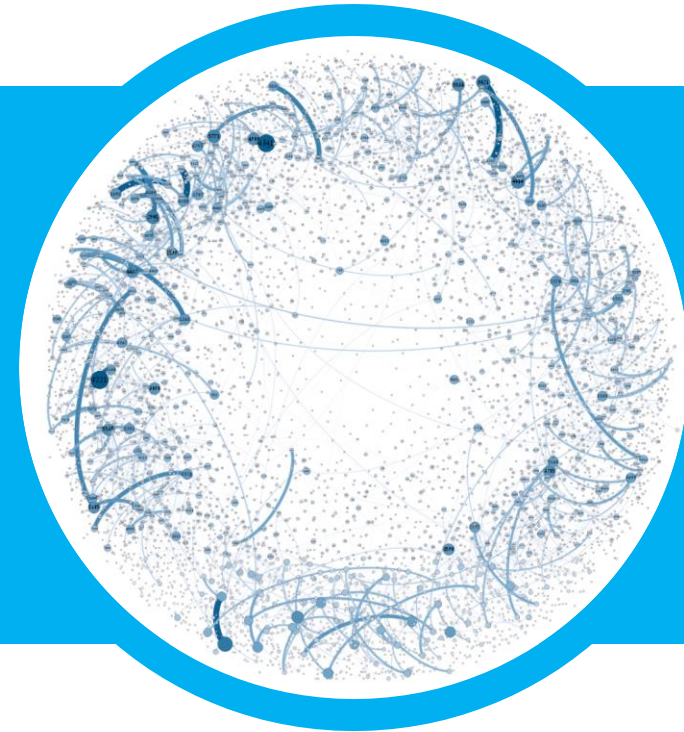


Network GPS: Navigating network dynamics

Baruch Barzel
Bar-Ilan University
www.barzellab.com



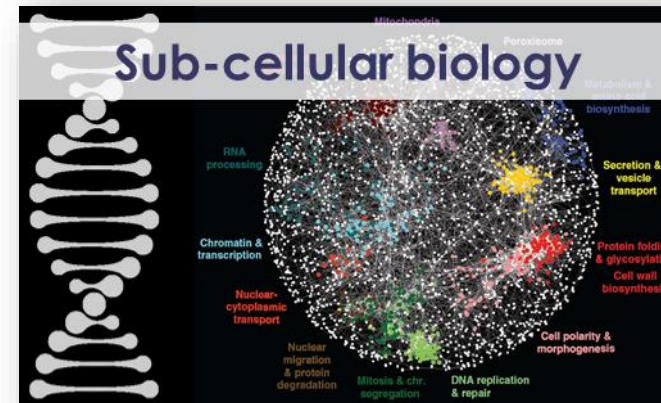
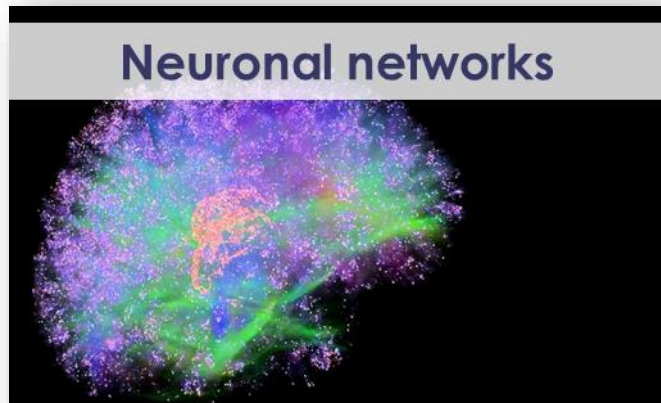
Going beyond mapping



Going beyond mapping



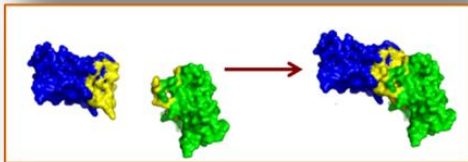
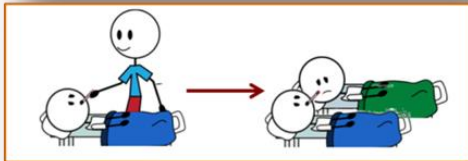
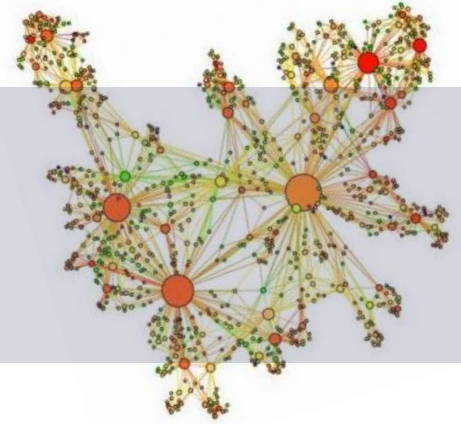
Network dynamics



Dynamics layer

$x_i(t) \rightarrow$ Activity

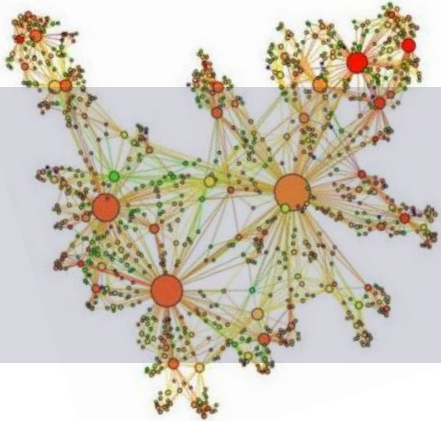
$$\frac{dx_i}{dt} = M_0(x_i) + \sum_{j=1}^N A_{ij} M_1(x_i) M_2(x_j)$$



A_{ij} Weighted, directed topology

M_0, M_1, M_2 Intrinsic nonlinear interaction mechanisms

Dynamics layer

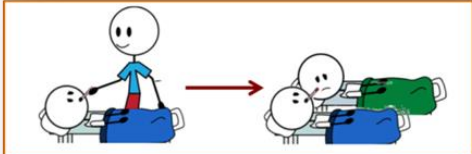


$x_i(t) \rightarrow$ Activity

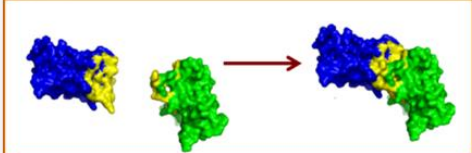
$$\frac{dx_i}{dt} = M_0(x_i) + \sum_{j=1}^N A_{ij} M_1(x_i) M_2(x_j)$$



$$\frac{dx_i}{dt} = Bx_i (1 - x_i) + \sum_{j=1}^N A_{ij} \frac{x_i x_j^a}{1 + x_j^a}$$

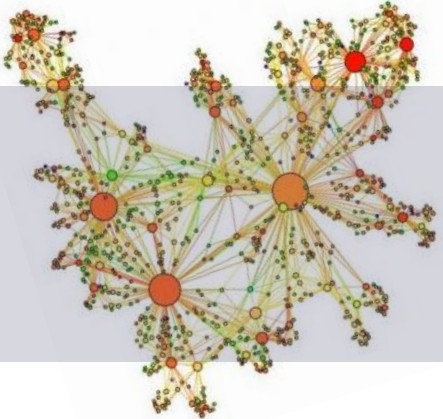


$$\frac{dx_i}{dt} = -Bx_i + \sum_{j=1}^N A_{ij} (1 - x_i) x_j$$



$$\frac{dx_i}{dt} = -Bx_i + \sum_{j=1}^N A_{ij} \frac{x_j^h}{1 + x_j^h}$$

Dynamics layer

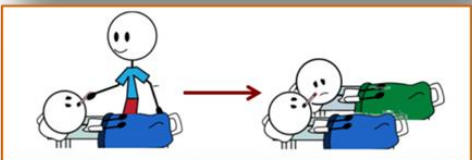


$x_i(t) \rightarrow$ Activity

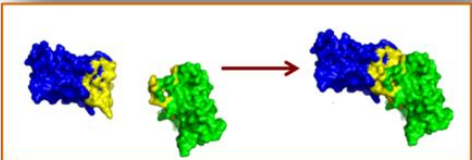
$$\frac{dx_i}{dt} = M_0(x_i) + \sum_{j=1}^N A_{ij} M_1(x_i) M_2(x_j)$$



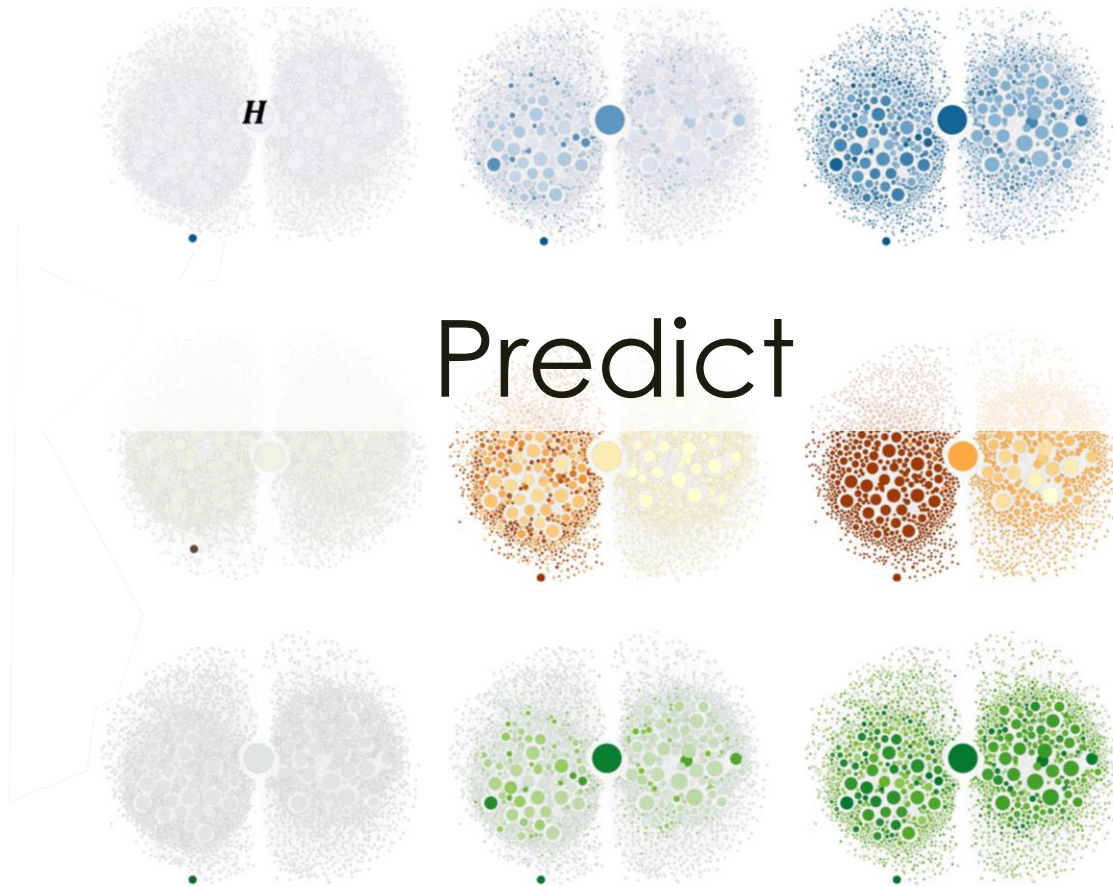
$$\frac{dx_i}{dt} = Bx_i (1 - x_i) + \sum_{j=1}^N A_{ij} \frac{x_i x_j^a}{1 + x_j^a}$$



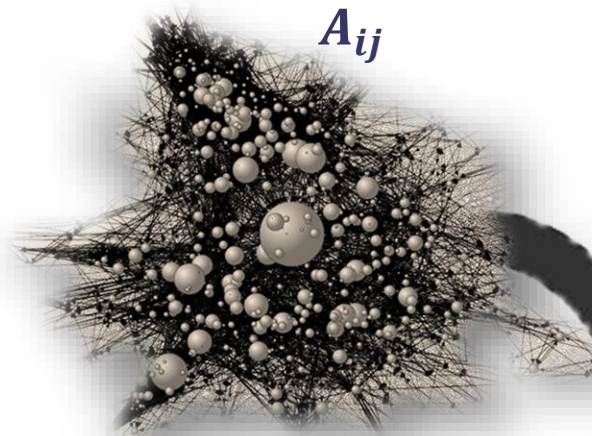
$$\frac{dx_i}{dt} = -Bx_i + \sum_{j=1}^N A_{ij} (1 - x_i) x_j$$



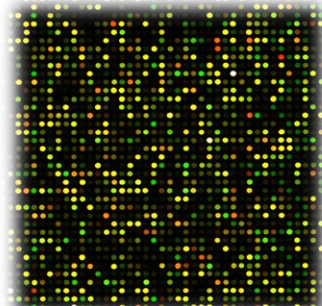
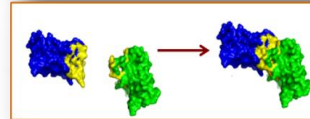
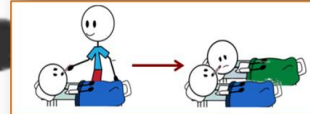
$$\frac{dx_i}{dt} = -Bx_i + \sum_{j=1}^N A_{ij} \frac{x_j^h}{1 + x_j^h}$$



Bringing networks to life



$M_0(x), M_1(x), M_2(x)$

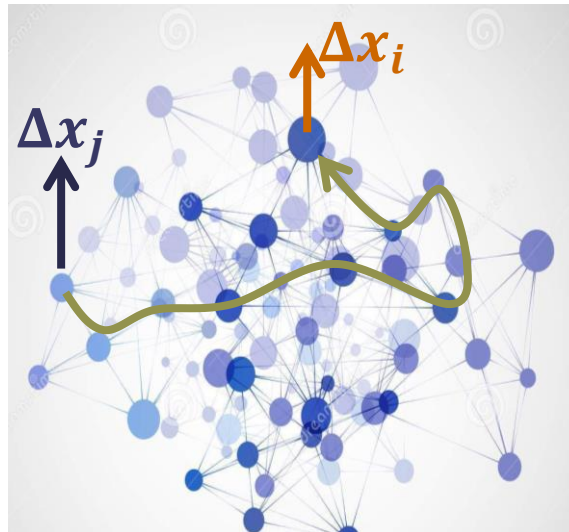


patterns of information spread

Information flow in complex networks

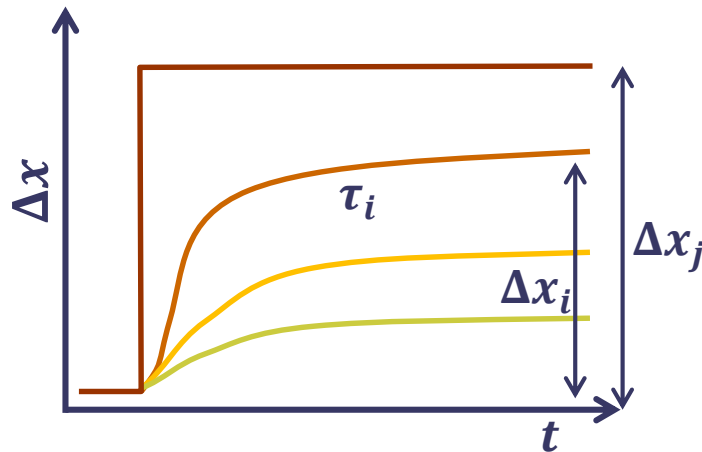
$$x_j \rightarrow x_j + \Delta x_j \longrightarrow x_i + \Delta x_i(t)$$

signal response



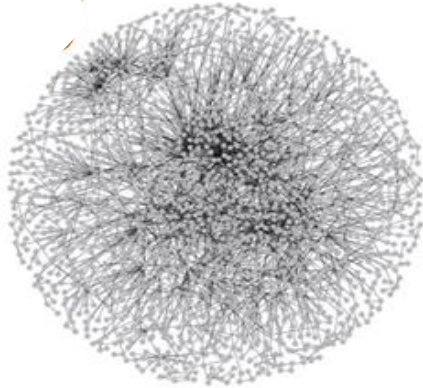
Information flow in complex networks

- When (τ_i)
- Where (l_{ij})
- How strongly (Δx_i)
- How ($\mathcal{F}_i, \mathcal{F}_{ij}$)



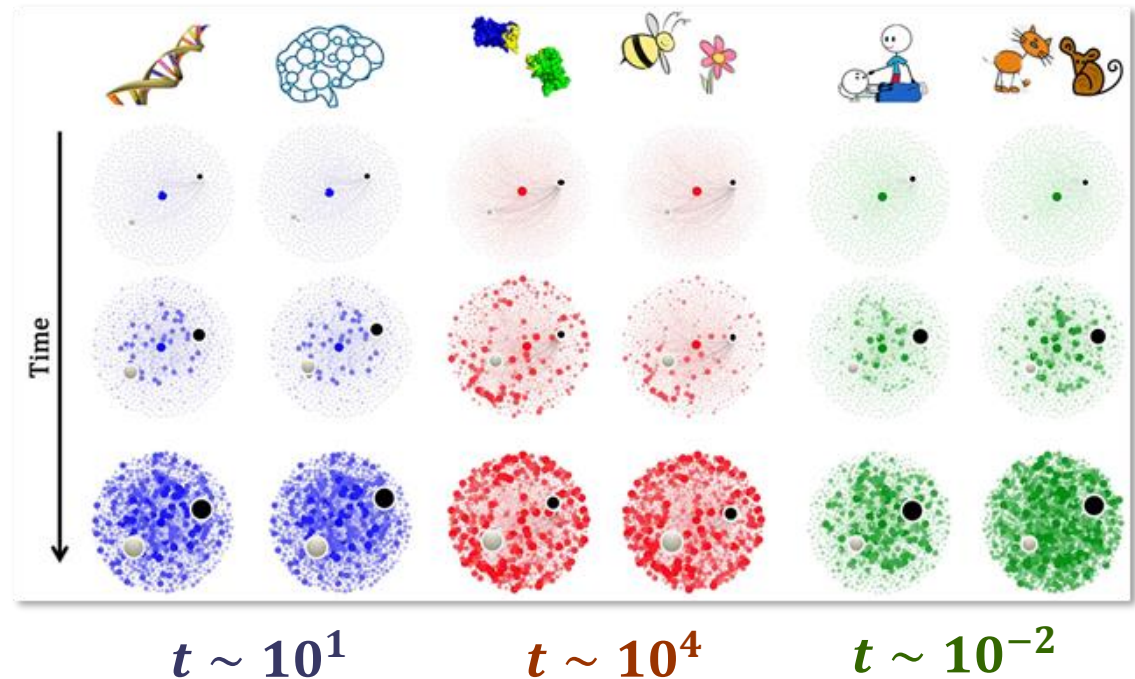
The zoo of propagation patterns

Topology

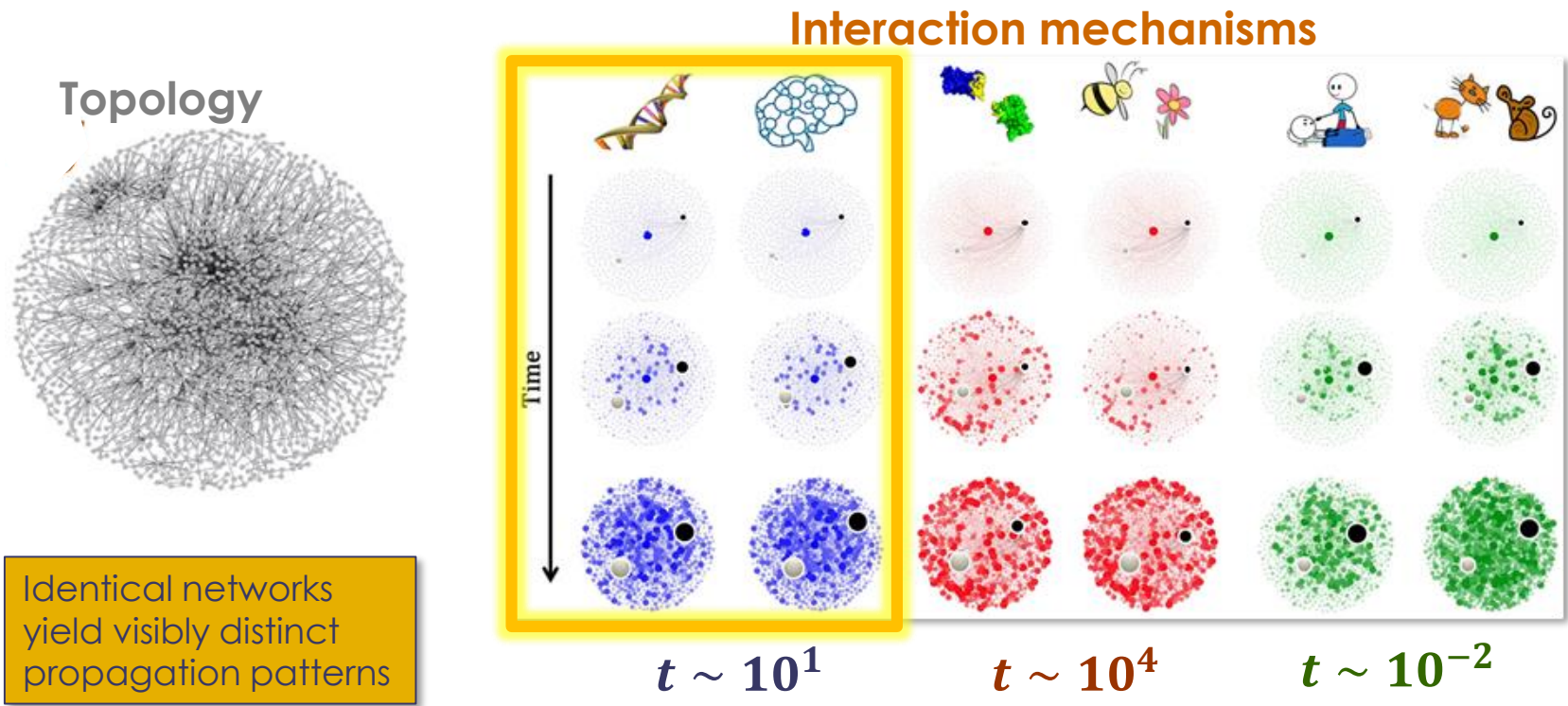


Identical networks
yield visibly distinct
propagation patterns

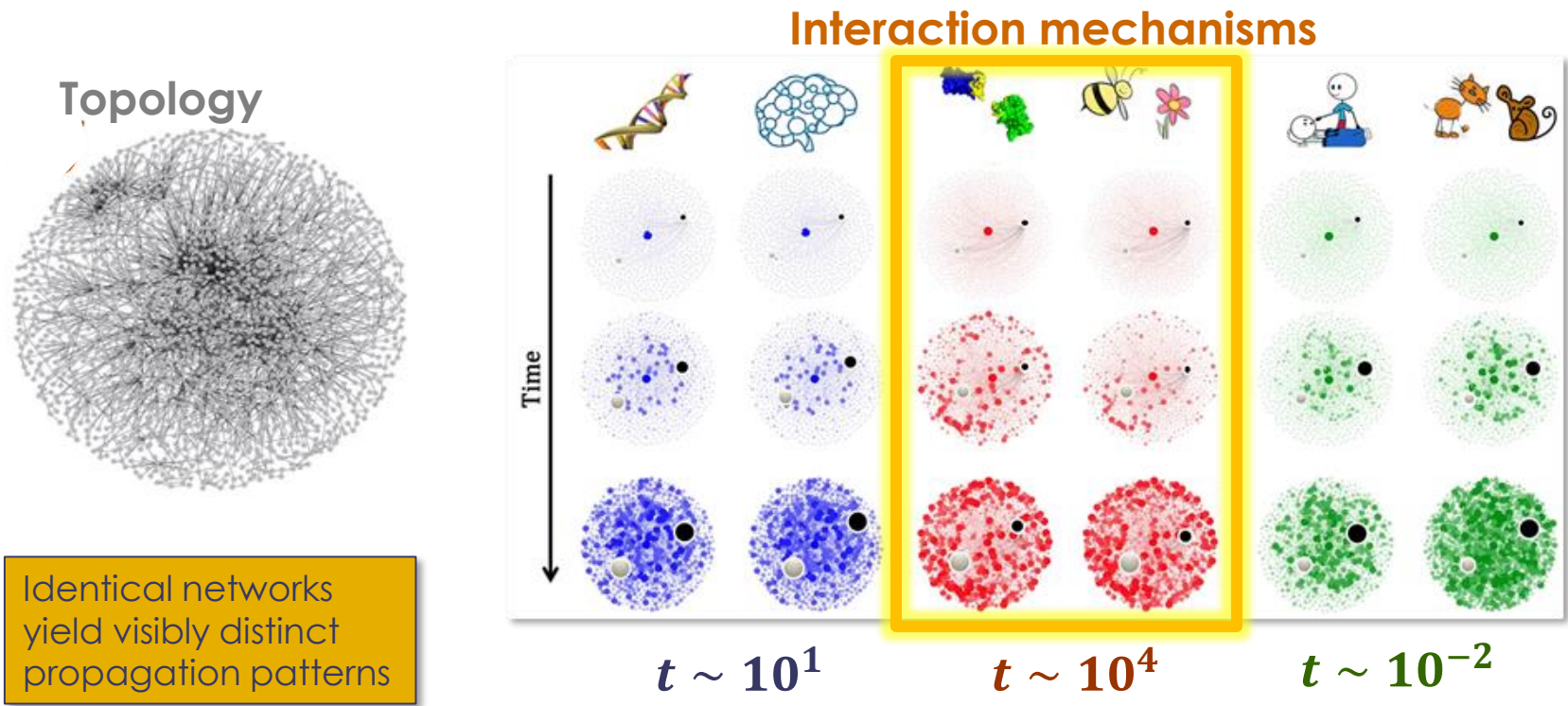
Interaction mechanisms



The zoo of propagation patterns

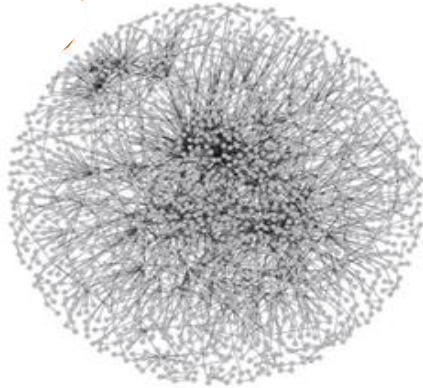


The zoo of propagation patterns



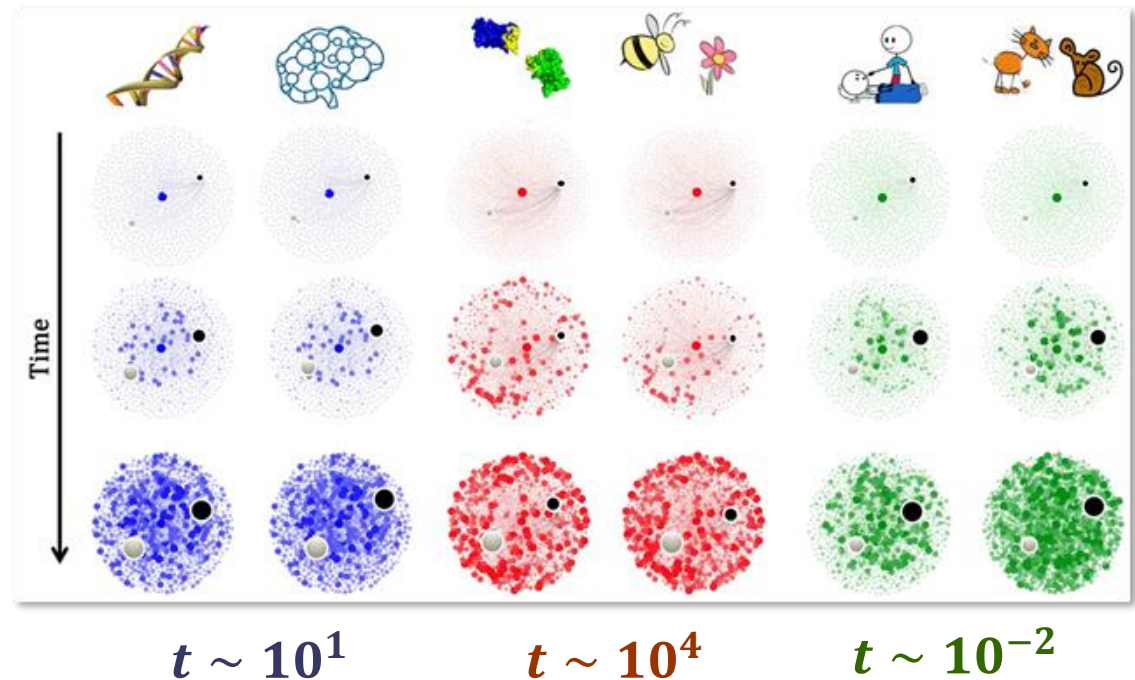
The zoo of propagation patterns

Topology



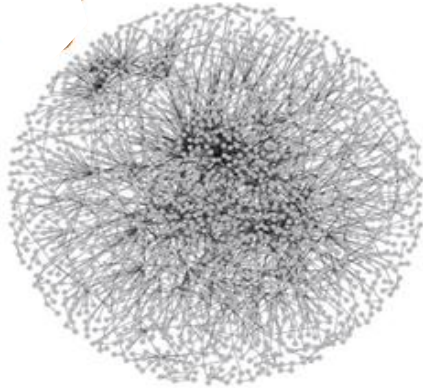
Identical networks
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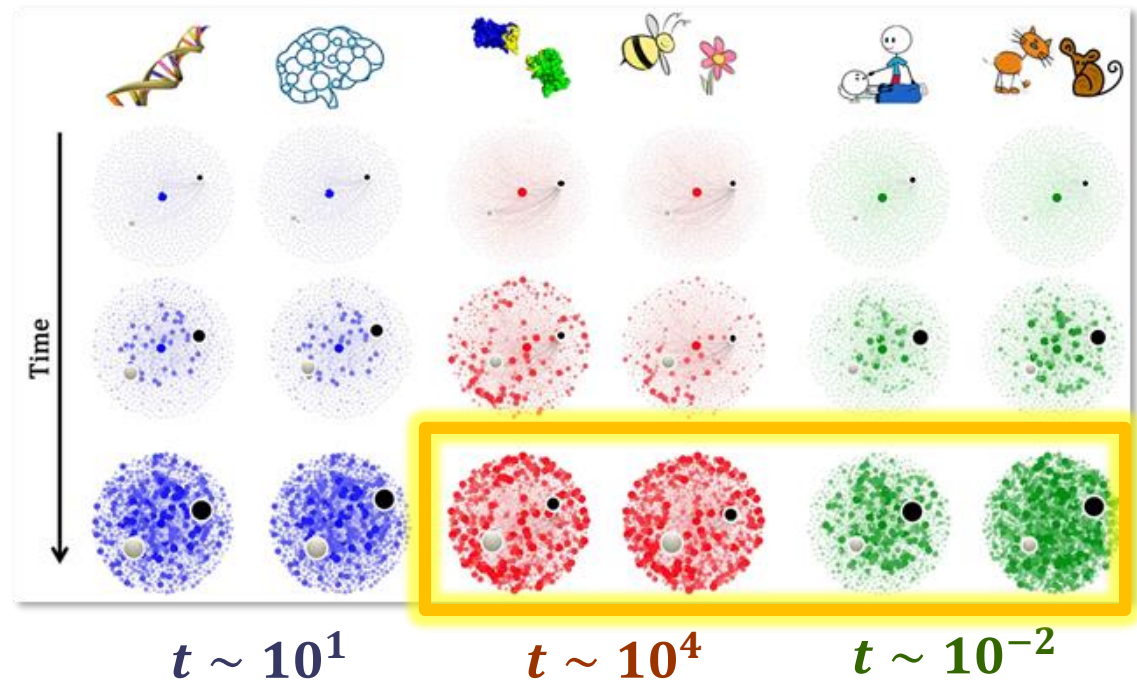
The zoo of propagation patterns

Topology

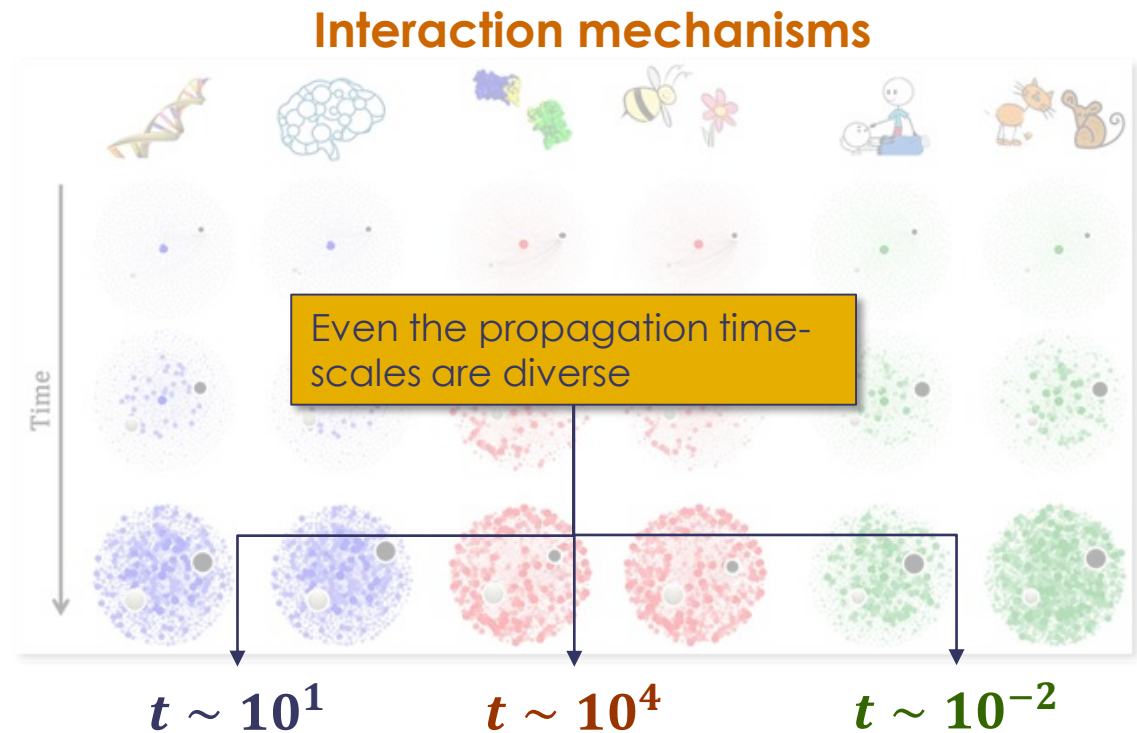
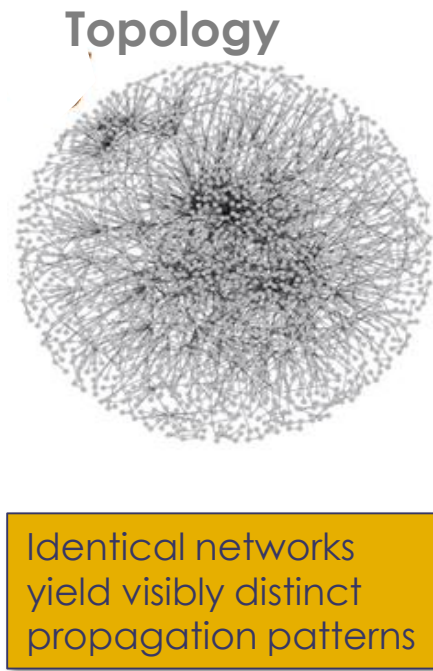


Identical networks yield visibly distinct propagation patterns

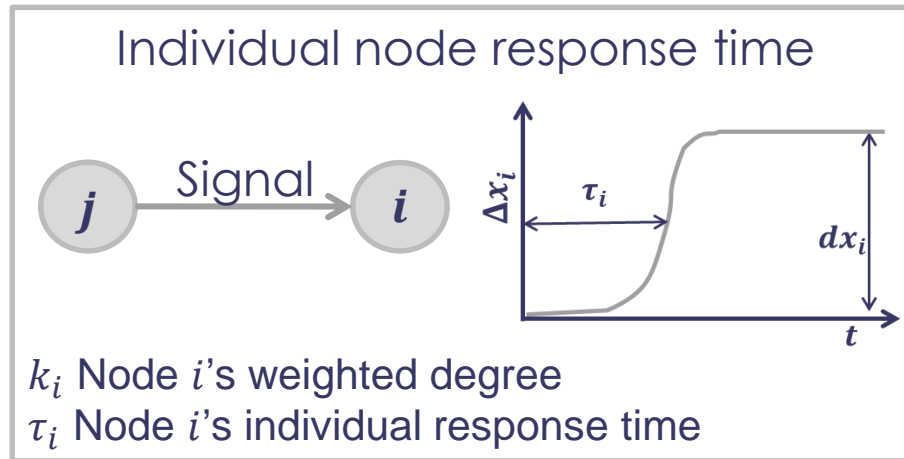
Interaction mechanisms



The zoo of propagation patterns



Taming the zoo of propagation patterns

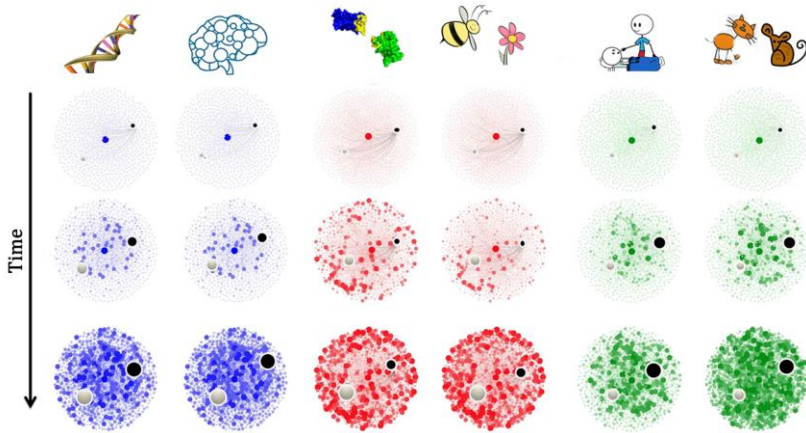


$$\tau_i \sim k_i^\theta$$

A node's intrinsic response time scales with its weighted degree

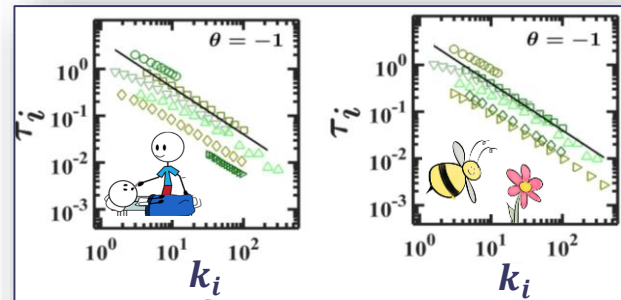
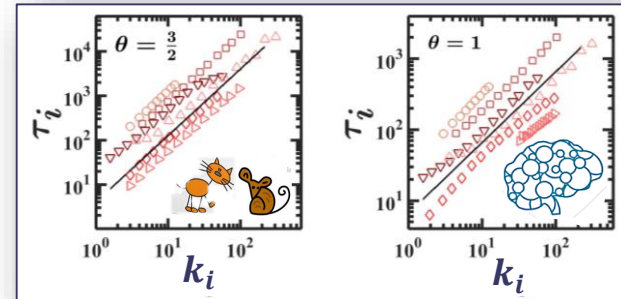
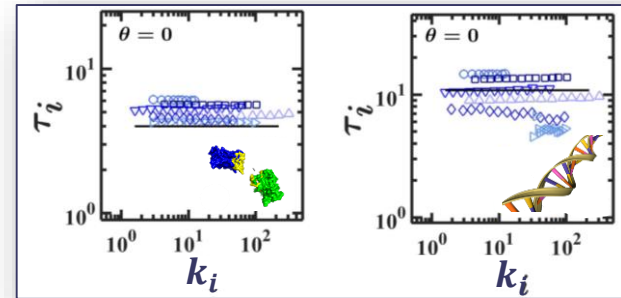
Why should you care?

Diverse & unpredictable



Universal

$$\tau_i \sim k_i^\theta$$



Dynamic insight

Response time

Dynamic observable of interest.

$$\tau_i \sim k_i^\theta$$

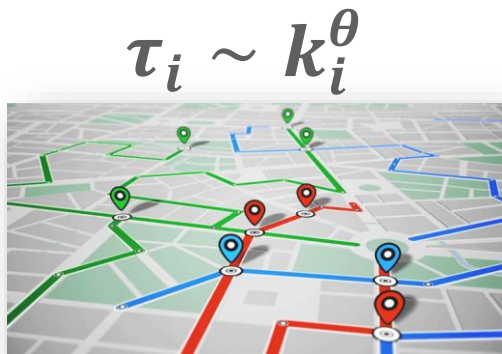
Weighted degree

Known topological element

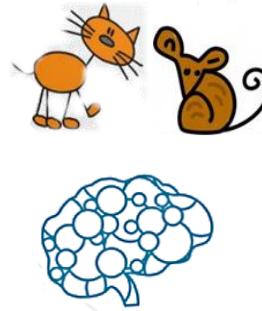
Dynamic determinant

Mapping topology into dynamics

Network GPS – how signals navigate the network



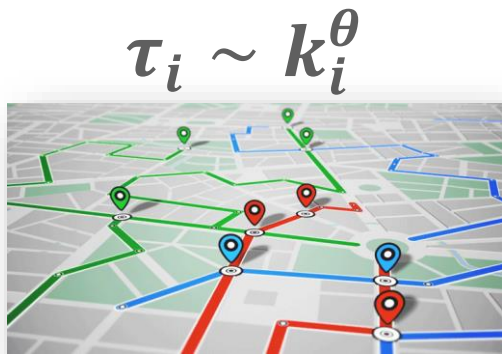
$\theta > 0$
Hubs
=
Bottlenecks



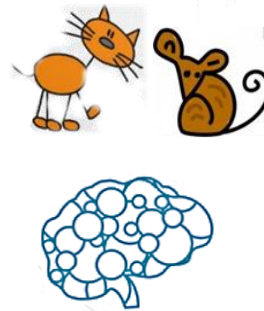
$\theta < 0$
Hubs
=
Free flow



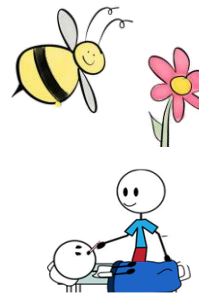
Network GPS – how signals navigate the network



$\theta > 0$
Hubs
= Bottlenecks

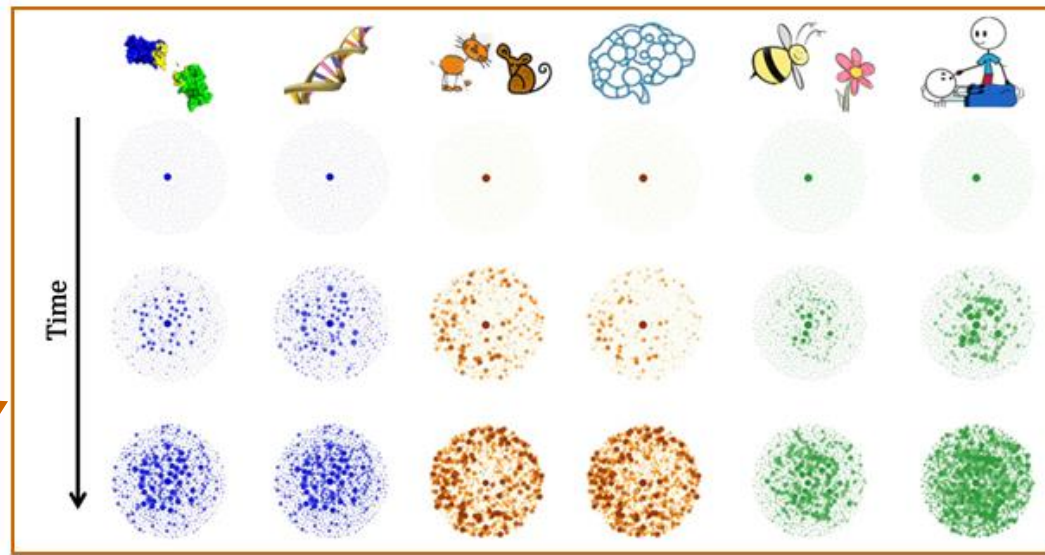
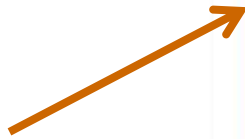


$\theta < 0$
Hubs
= Free flow



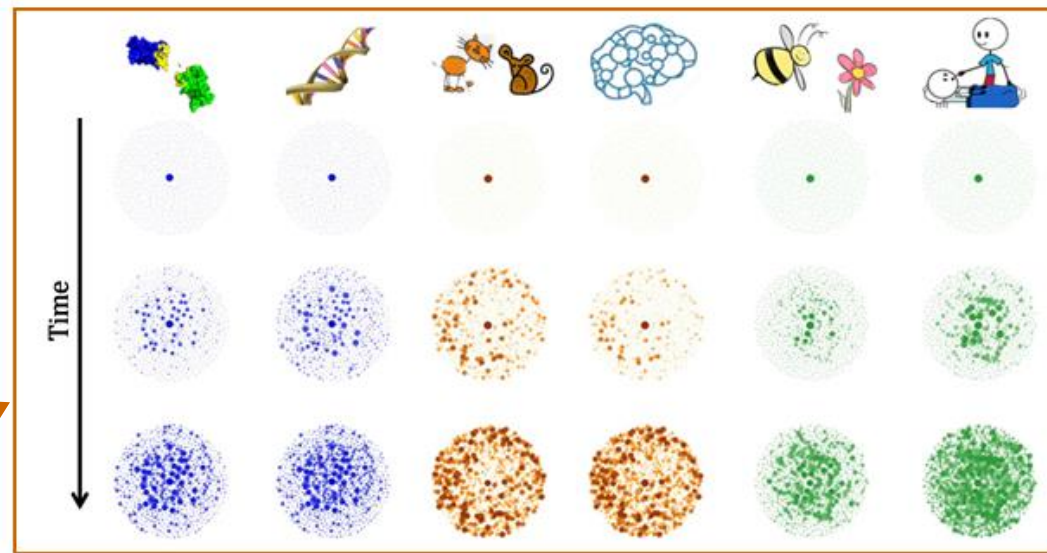
Network GPS – how signals navigate the network

Naïve layout

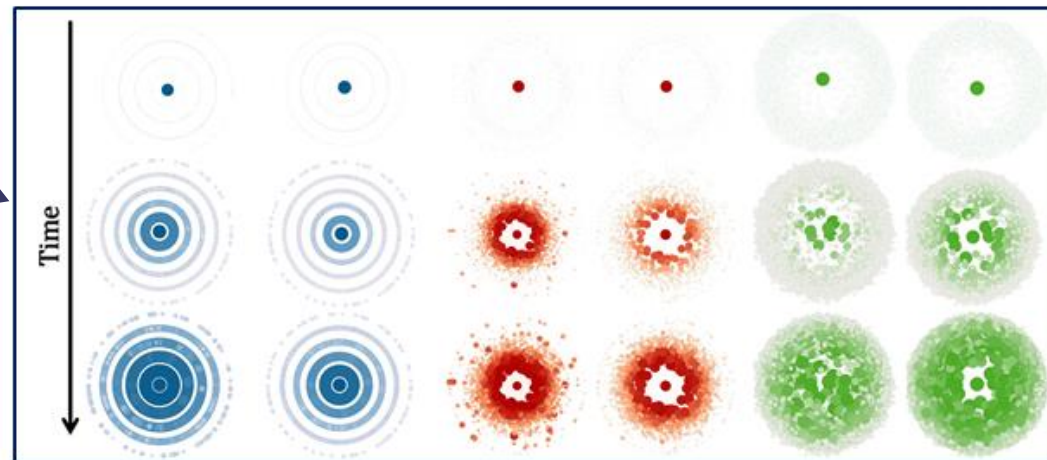


Network GPS – how signals navigate the network

Naïve layout

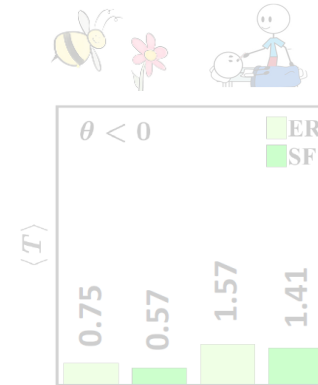
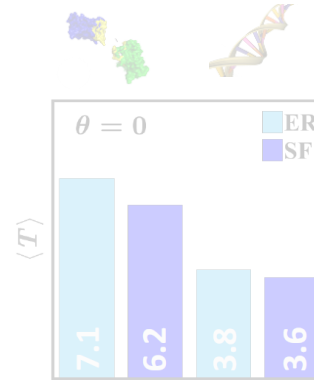
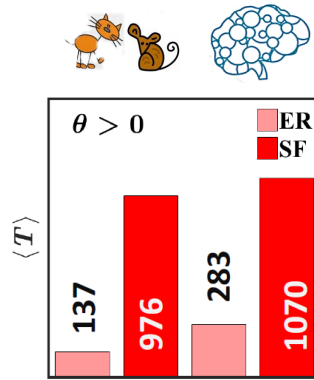


Our GPS prediction



Same topology – different spreading rules

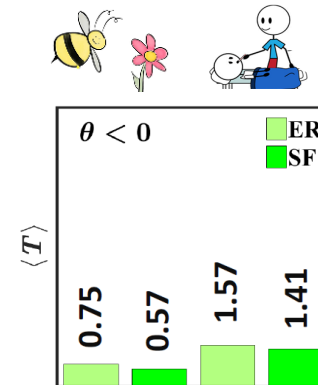
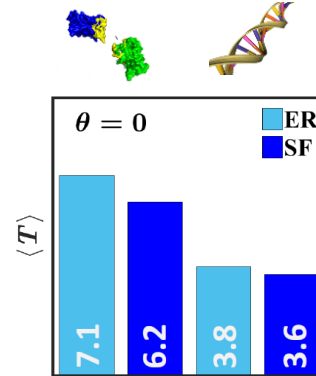
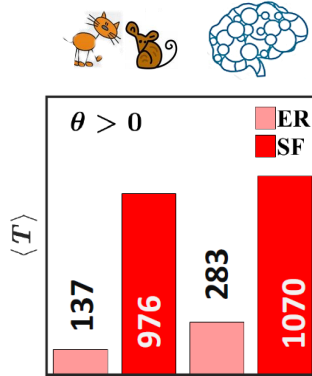
$\theta > 0$
Hubs dramatically slow down propagation



Hubs speed up the propagation

Same topology – different spreading rules

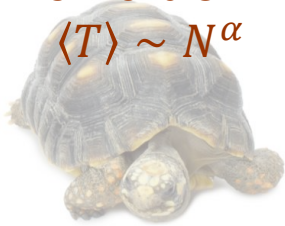
$\theta > 0$
Hubs
dramatically
slow down
propagation



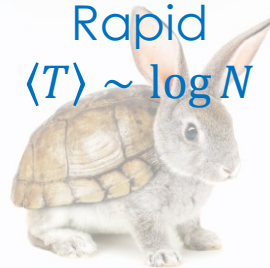
$\theta < 0$
Hubs speed up
the
propagation



Ultra-slow
 $\langle T \rangle \sim N^\alpha$



Rapid
 $\langle T \rangle \sim \log N$

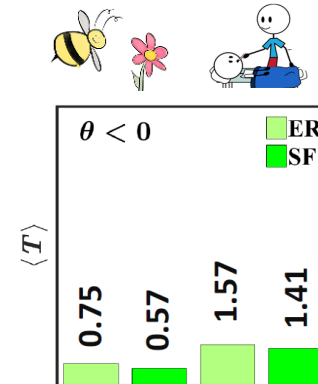
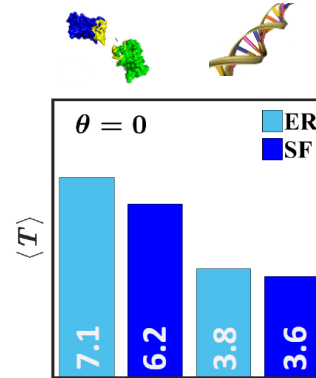
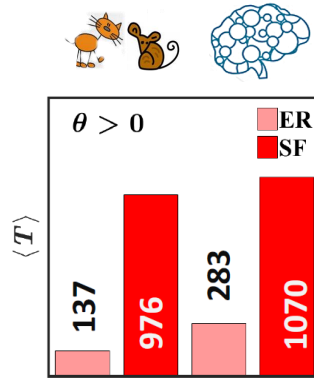


Ultra-fast
 $\langle T \rangle \sim \text{Const}$



Same topology – different spreading rules

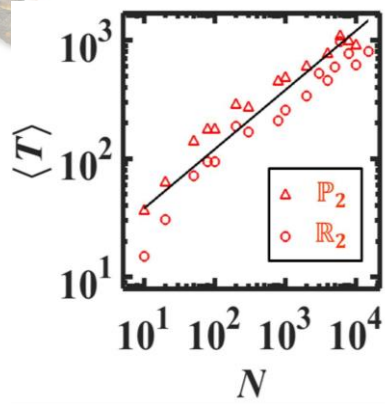
$\theta > 0$
Hubs dramatically slow down propagation



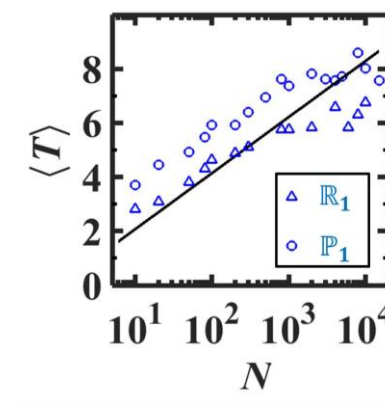
$\theta < 0$
Hubs speed up the propagation



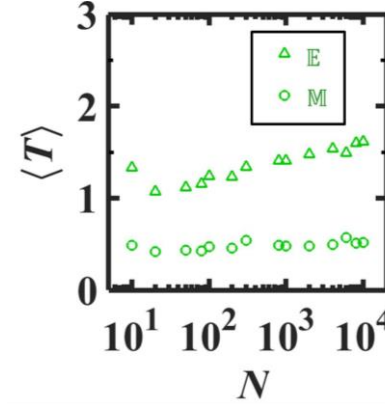
Ultra-slow
 $\langle T \rangle \sim N^\alpha$



Rapid
 $\langle T \rangle \sim \log N$

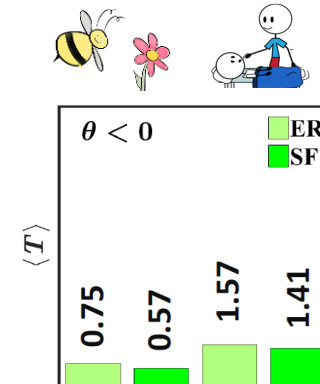
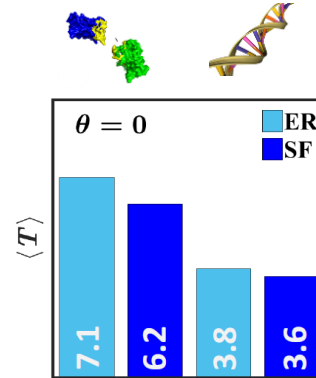
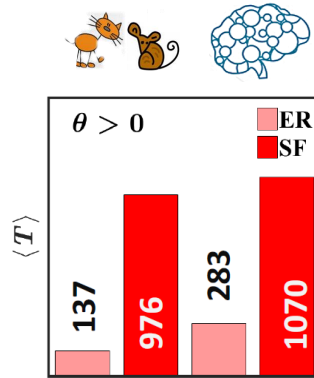


Ultra-fast
 $\langle T \rangle \sim \text{Const}$



Same topology – different spreading rules

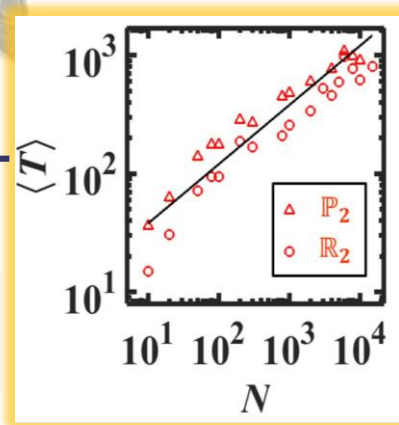
$\theta > 0$
Hubs dramatically slow down propagation



$\theta < 0$
Hubs speed up the propagation

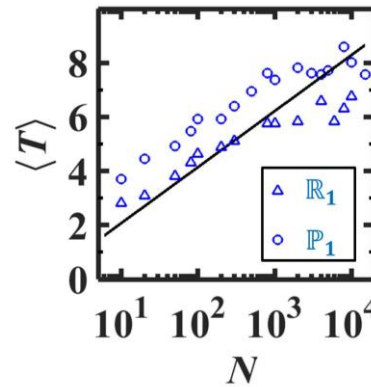


Ultra-slow
 $\langle T \rangle \sim N^\alpha$

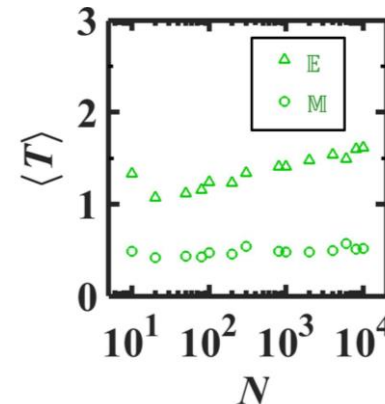


Lattice-like propagation

Rapid
 $\langle T \rangle \sim \log N$

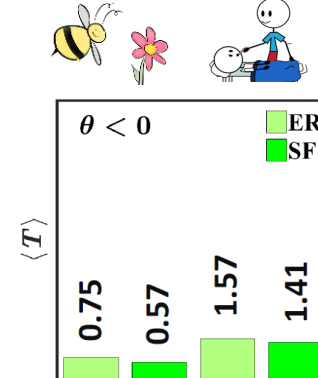
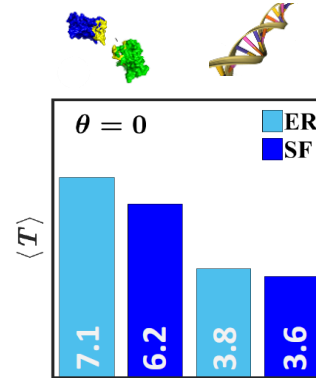
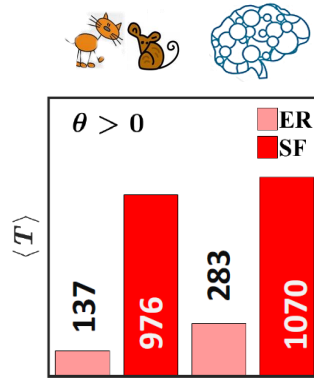


Ultra-fast
 $\langle T \rangle \sim \text{Const}$



Same topology – different spreading rules

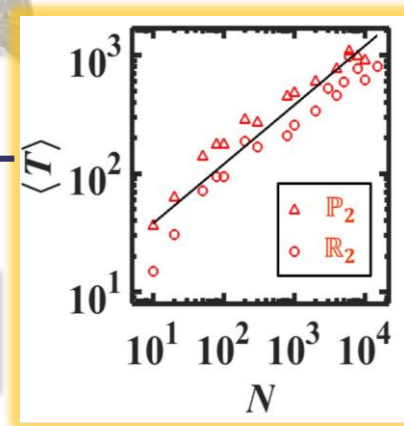
$\theta > 0$
Hubs dramatically slow down propagation



$\theta < 0$
Hubs speed up the propagation

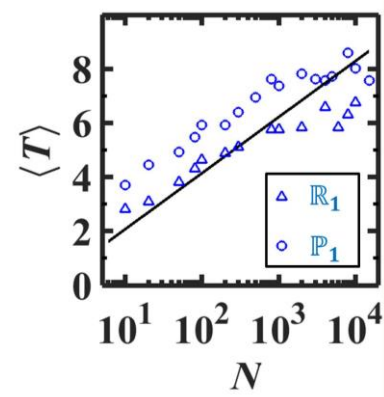


Ultra-slow
 $\langle T \rangle \sim N^\alpha$

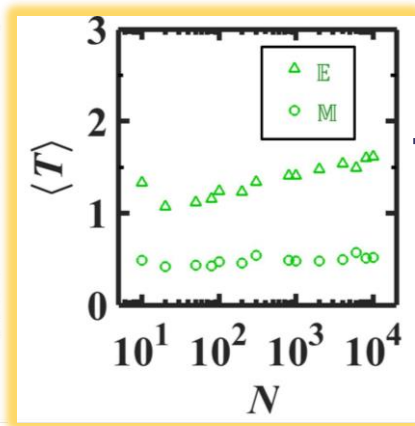


Lattice-like propagation

Rapid
 $\langle T \rangle \sim \log N$



Ultra-fast
 $\langle T \rangle \sim \text{Const}$



Single-node propagation

Same topology – different spreading rules

Different interpretations of scale-freeness. All boils down to a single analytically predictable parameter θ



Ultra-slow
 $\langle T \rangle \sim N^\alpha$

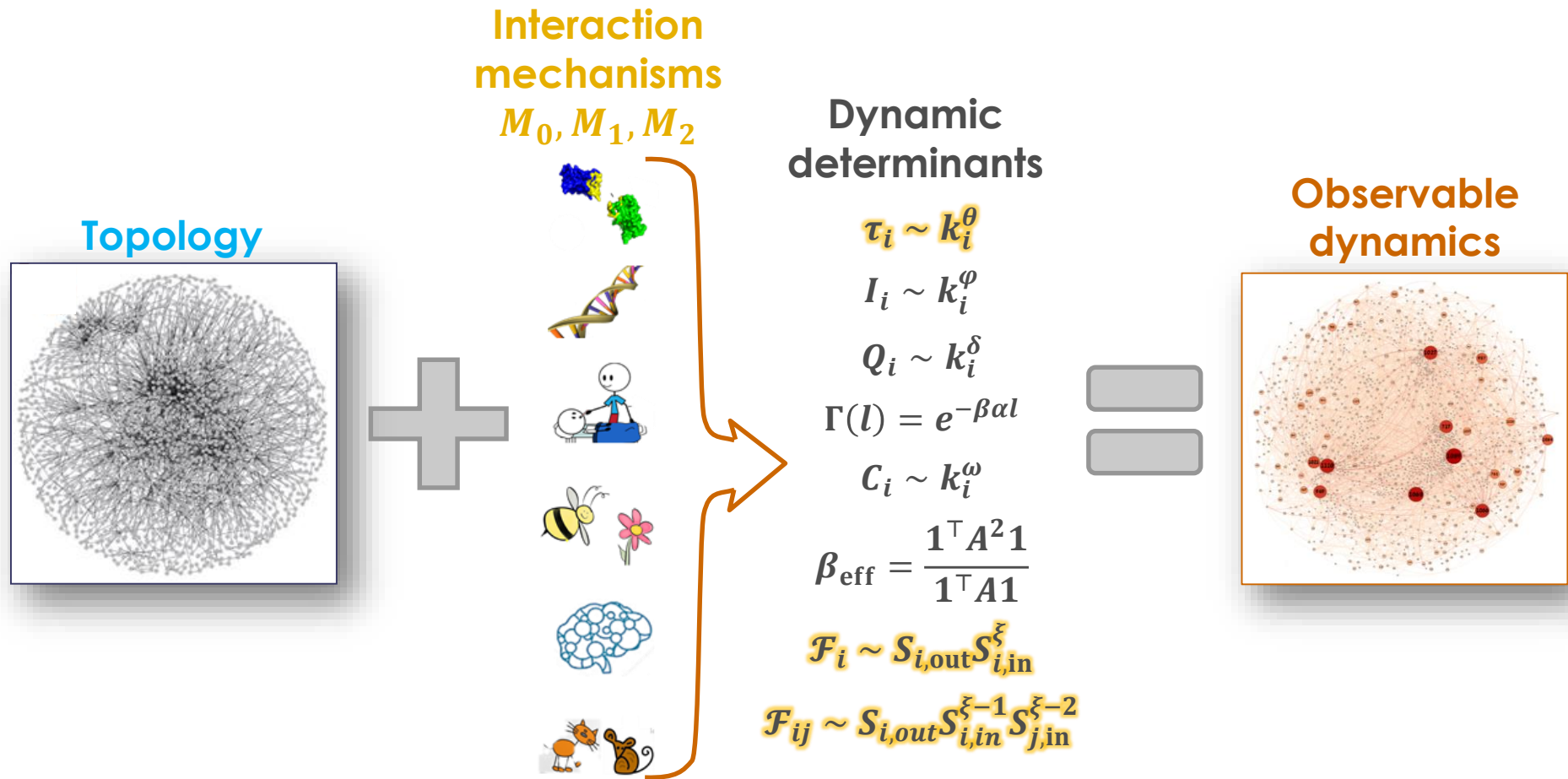


Rapid
 $\langle T \rangle \sim \log N$

Ultra-fast
 $\langle T \rangle \sim \text{Const}$

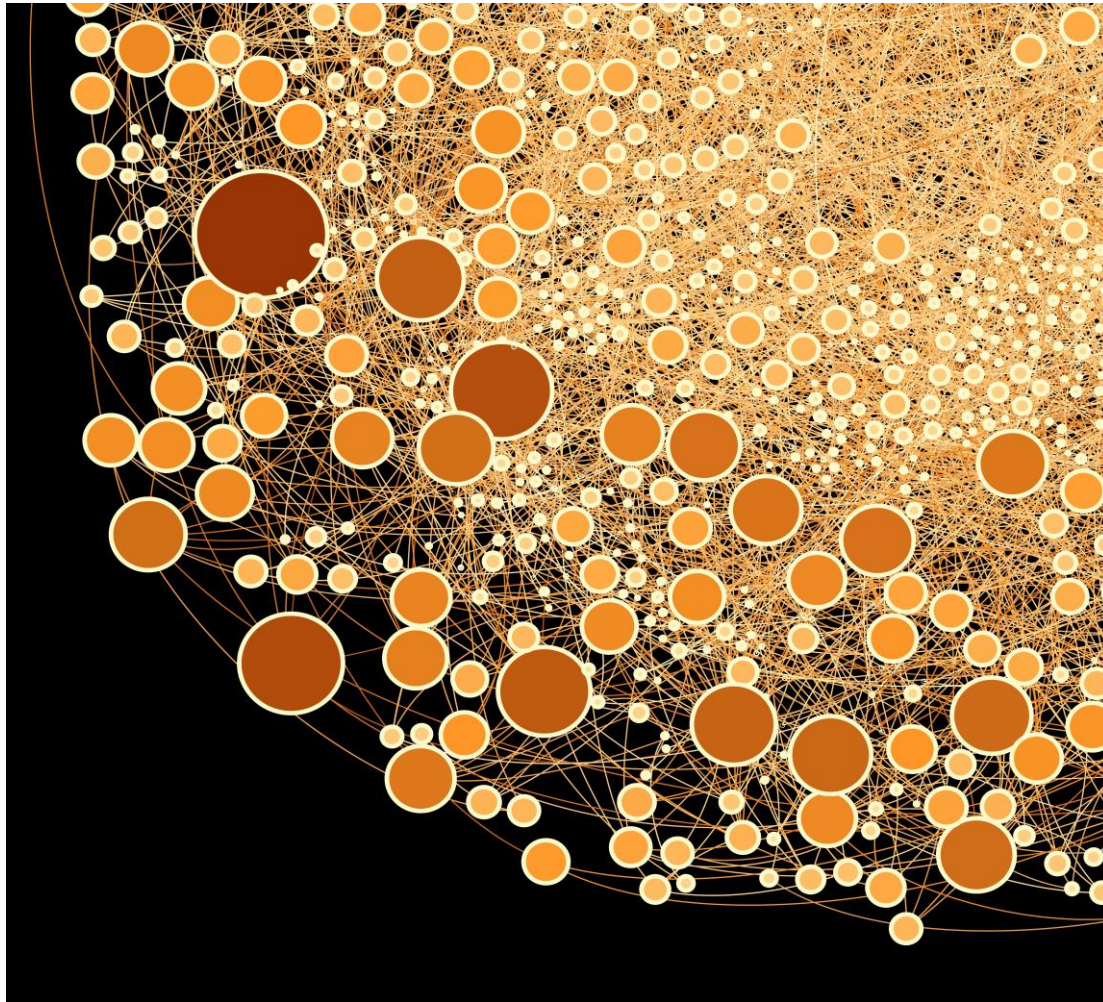


Dictionary of network dynamics

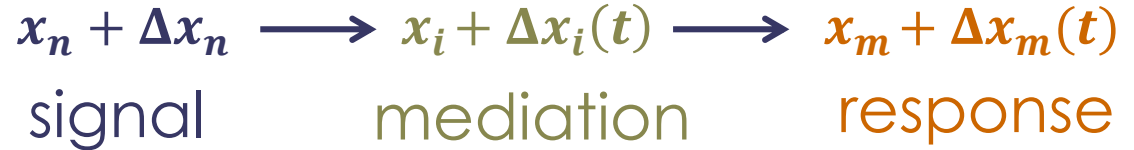
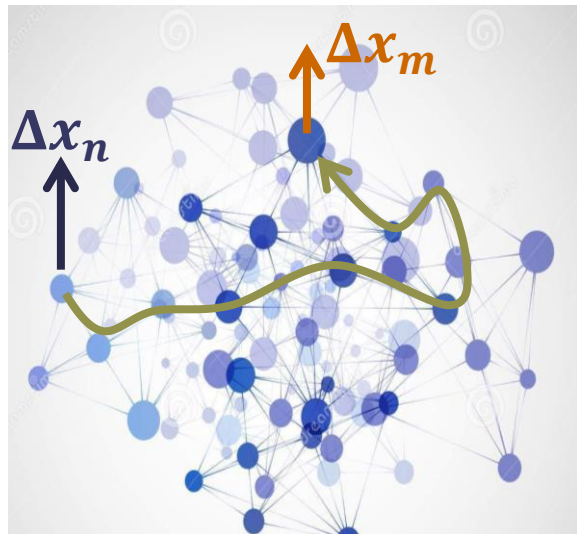


Microscopic Diversity condenses into a discrete set of
Universality classes that determine how
Topology translates into observable Dynamics

Understand



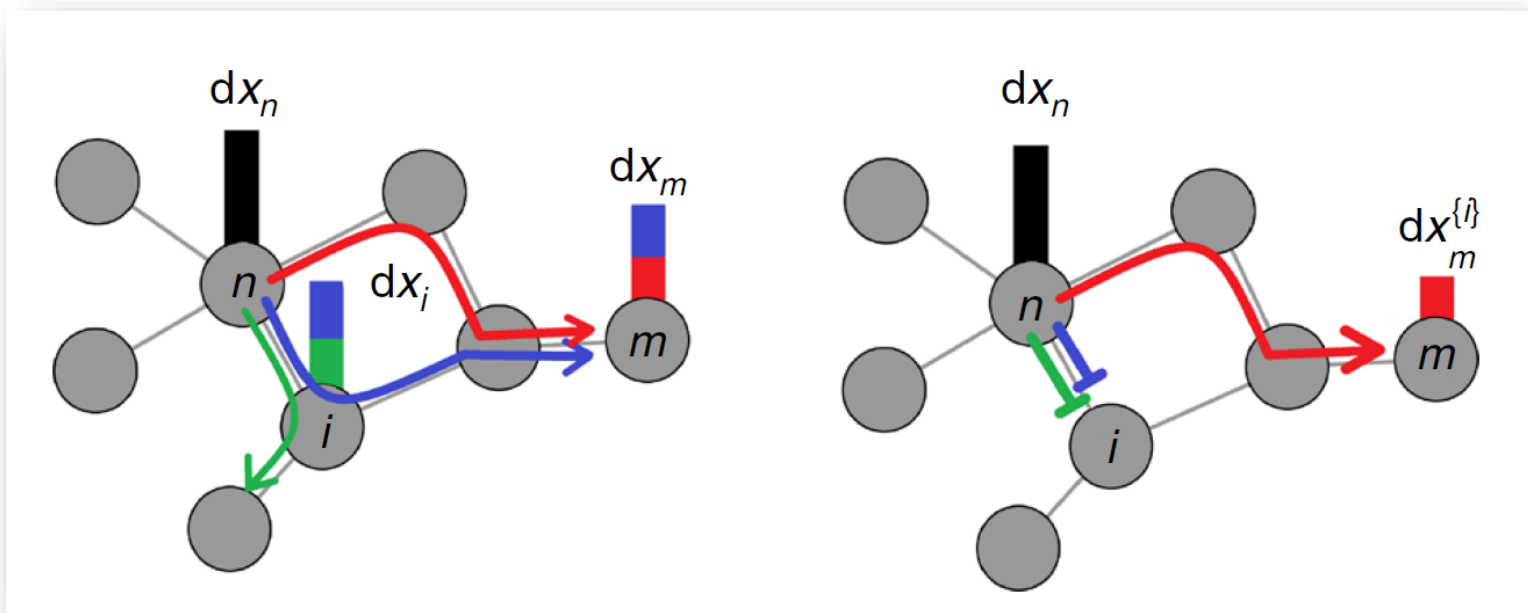
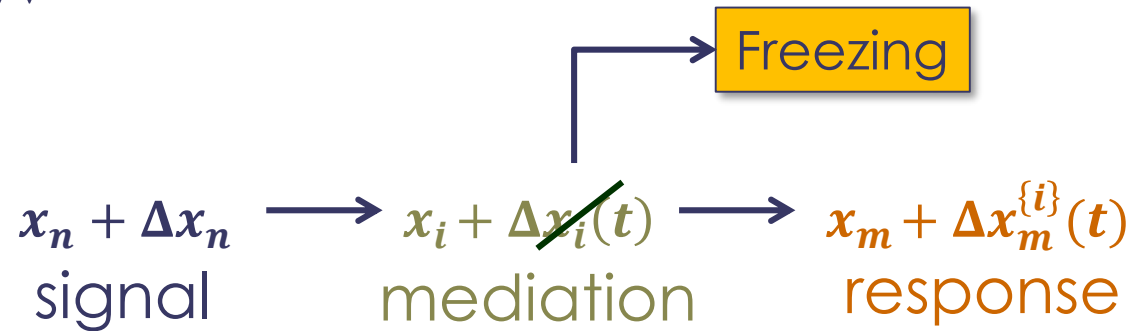
Information flow



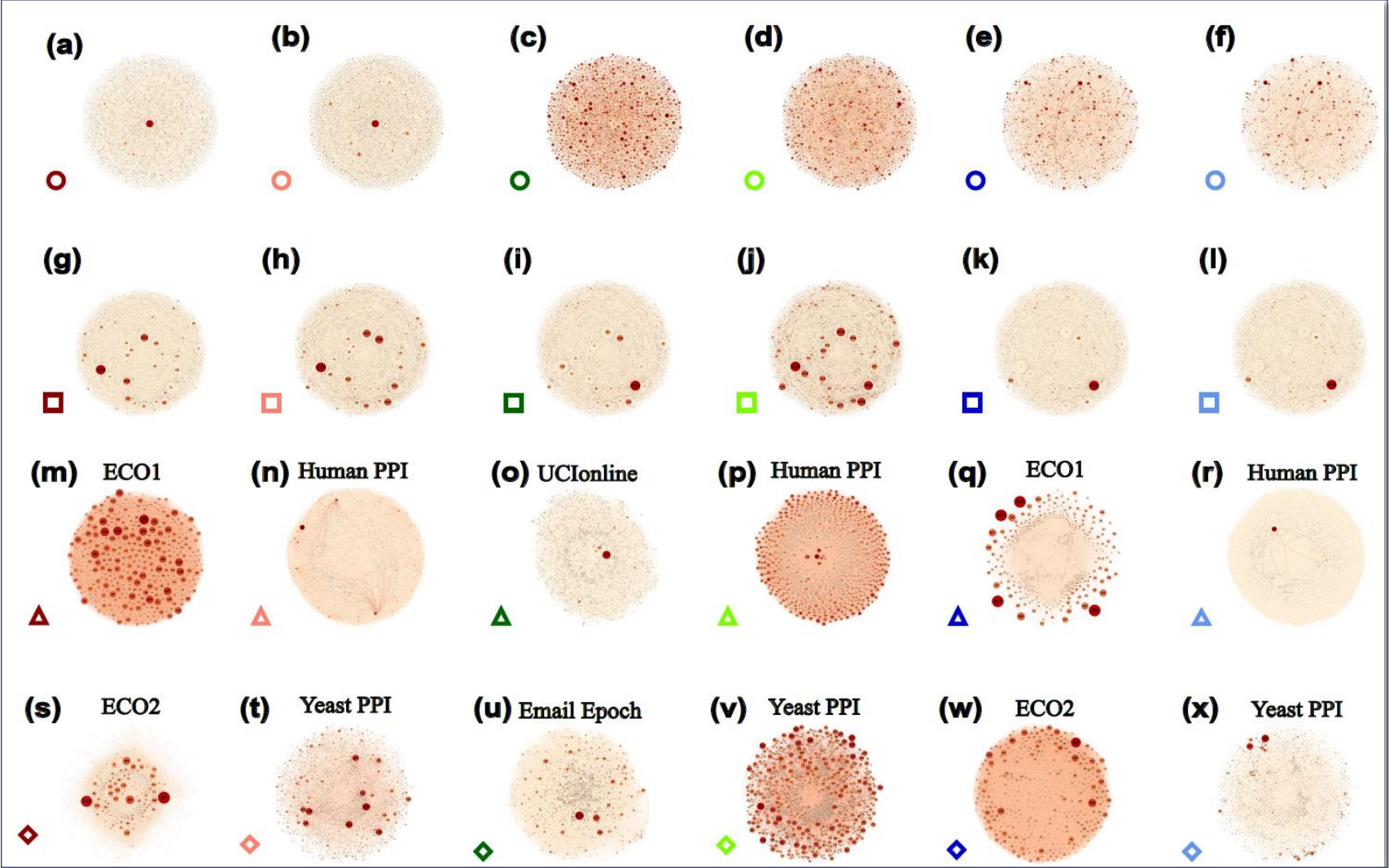
Node flow \mathcal{F}_i : how effective is each node in transferring information?

Link flow \mathcal{F}_{ij} : How effective is each link, pathway?

Information flow



The zoo of information flow patterns



Taming the zoo of information flow patterns

$$\mathcal{F}_i \sim k_{i,out} k_{i,in}^{\omega-1}$$

$$\mathcal{F}_{ij} \sim A_{ij} k_{i,out} k_{i,in}^{\xi-1} k_{j,in}^{\xi}$$

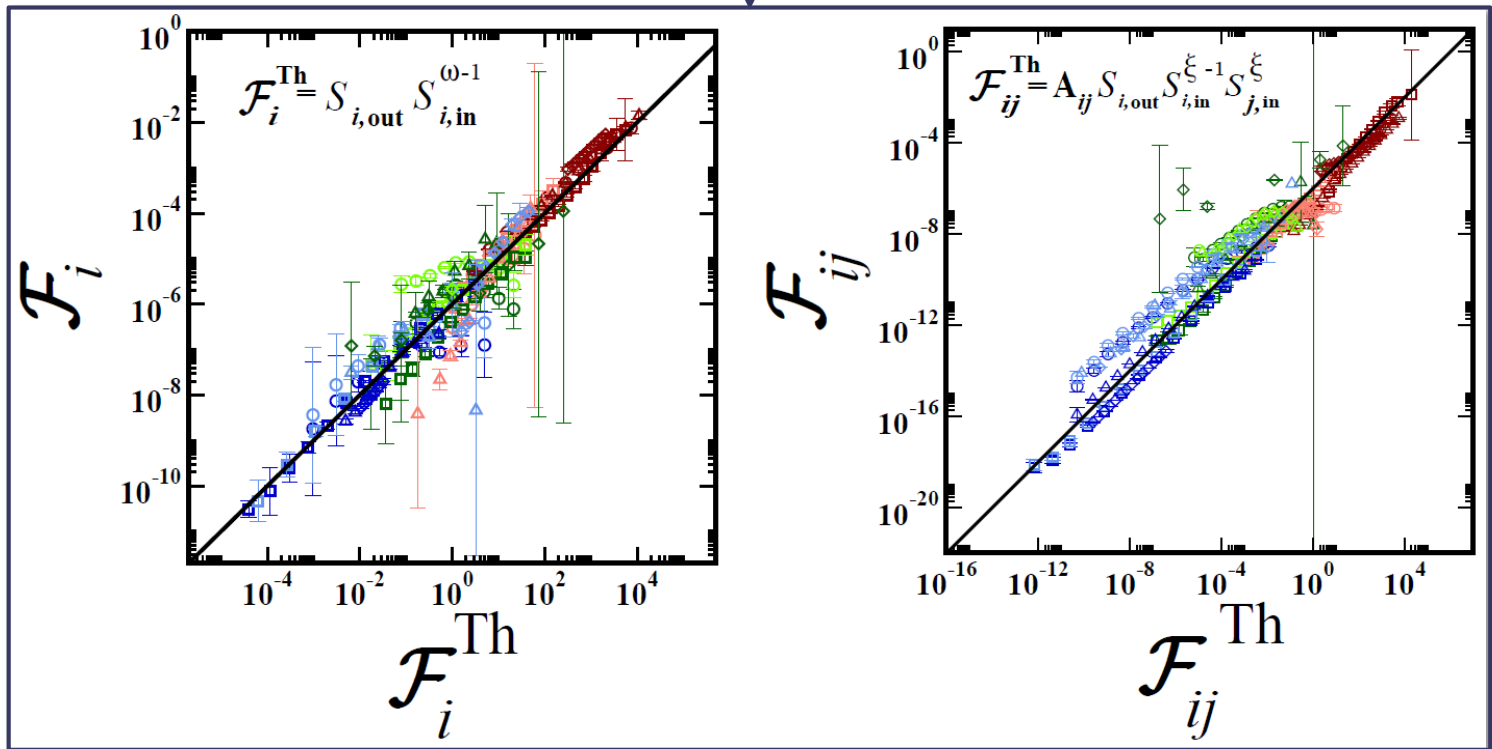
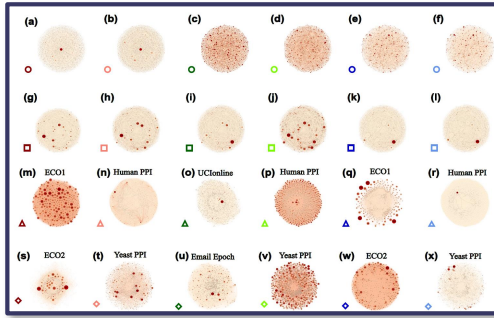
Node/link flow
Dynamic function
of interest.

**Dynamic
determinants**
Mapping topology into
dynamics



Weighted in/out degrees
Known topological elements

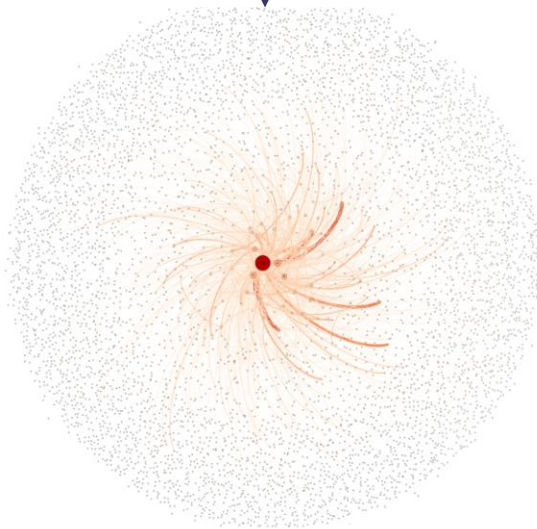
The ~~zoo~~ universal information flow patterns



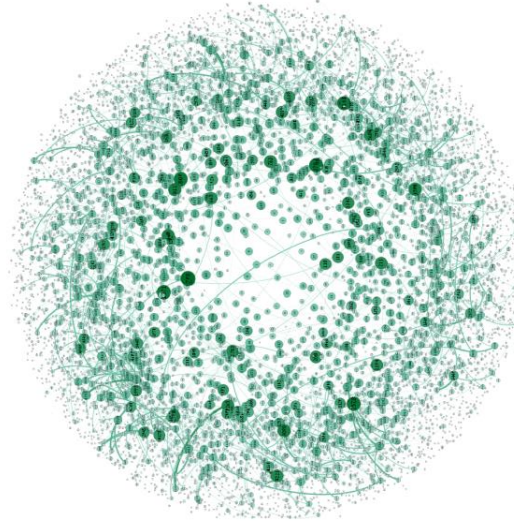
Universal classes of information flow patterns

$$\mathcal{F}_i \sim k_{i,\text{out}} k_{i,\text{in}}^{\omega-1}$$

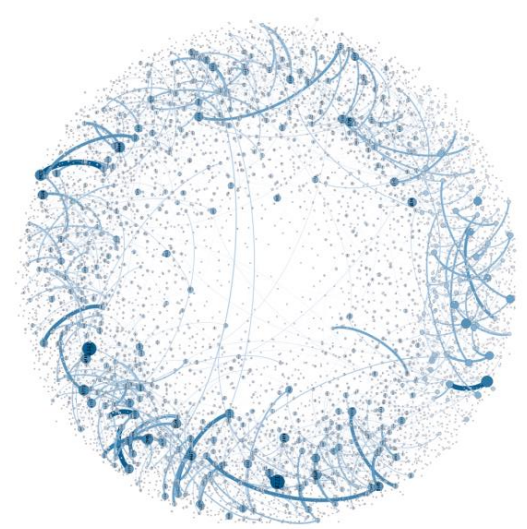
$$\mathcal{F}_{ij} \sim A_{ij} k_{i,\text{out}} k_{i,\text{in}}^{\xi-1} k_{j,\text{in}}^{\xi}$$



Hub centric

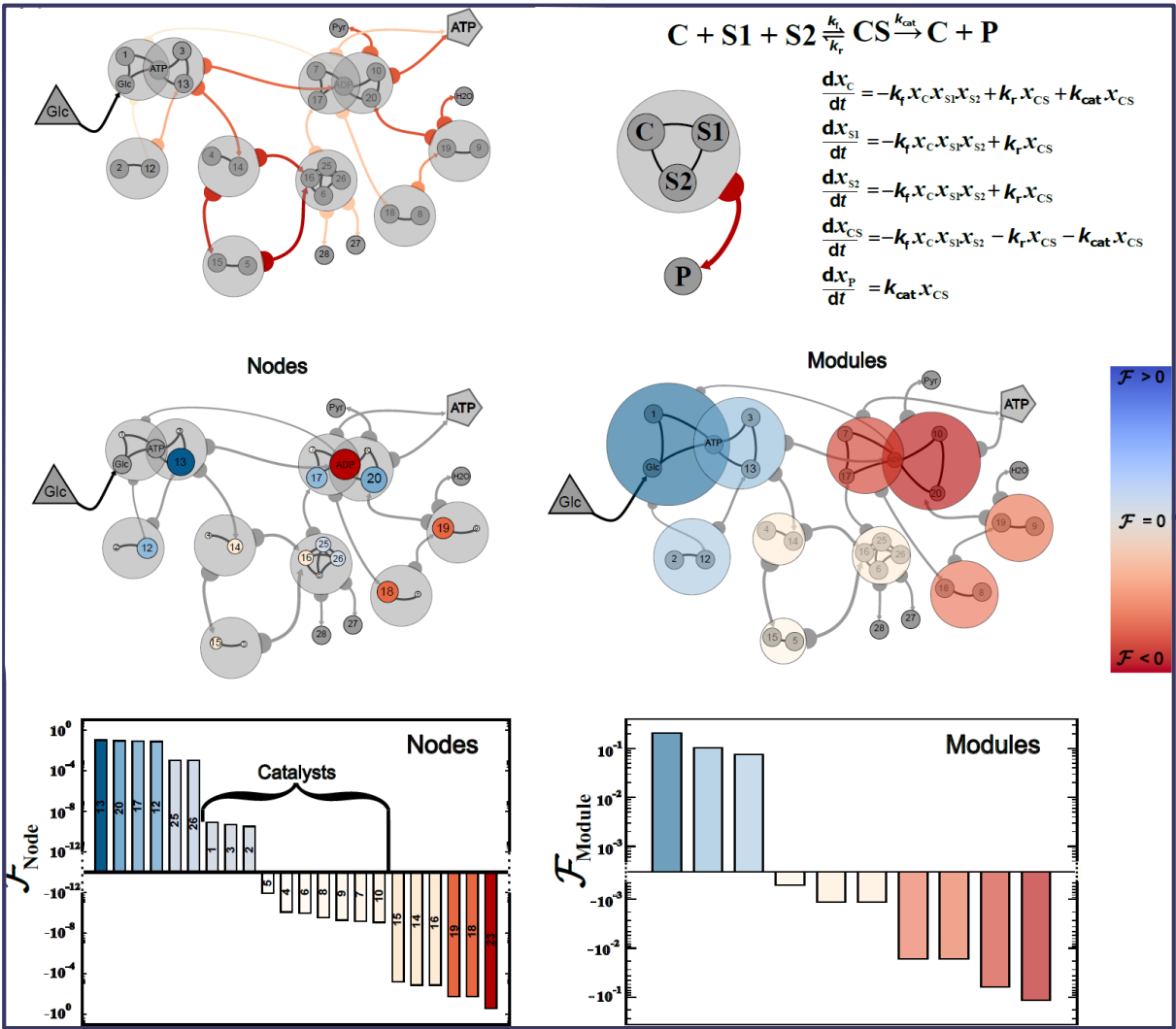


Egalitarian



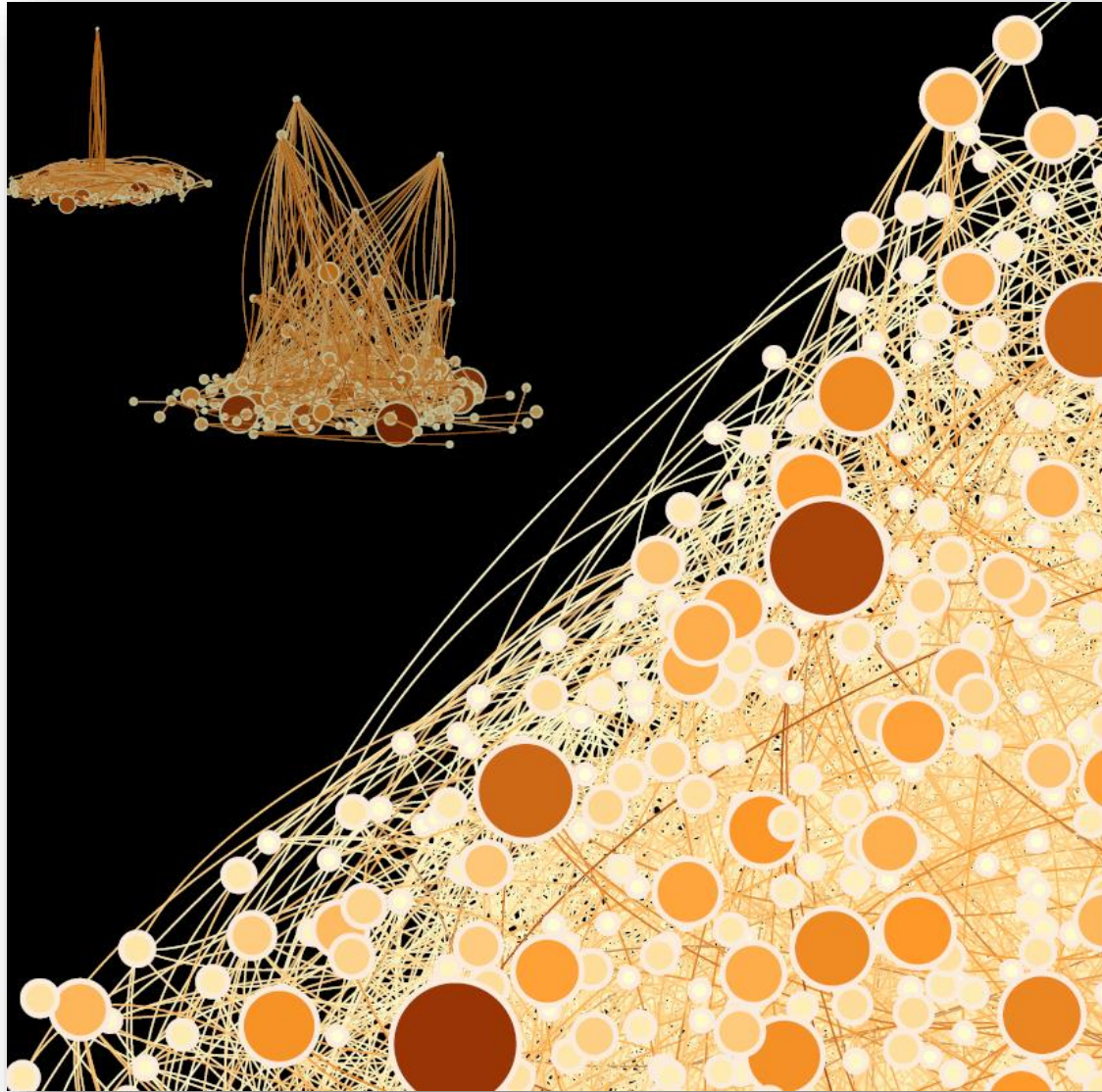
Peripheral

Glycolysis: do cells prefer traffic jams?



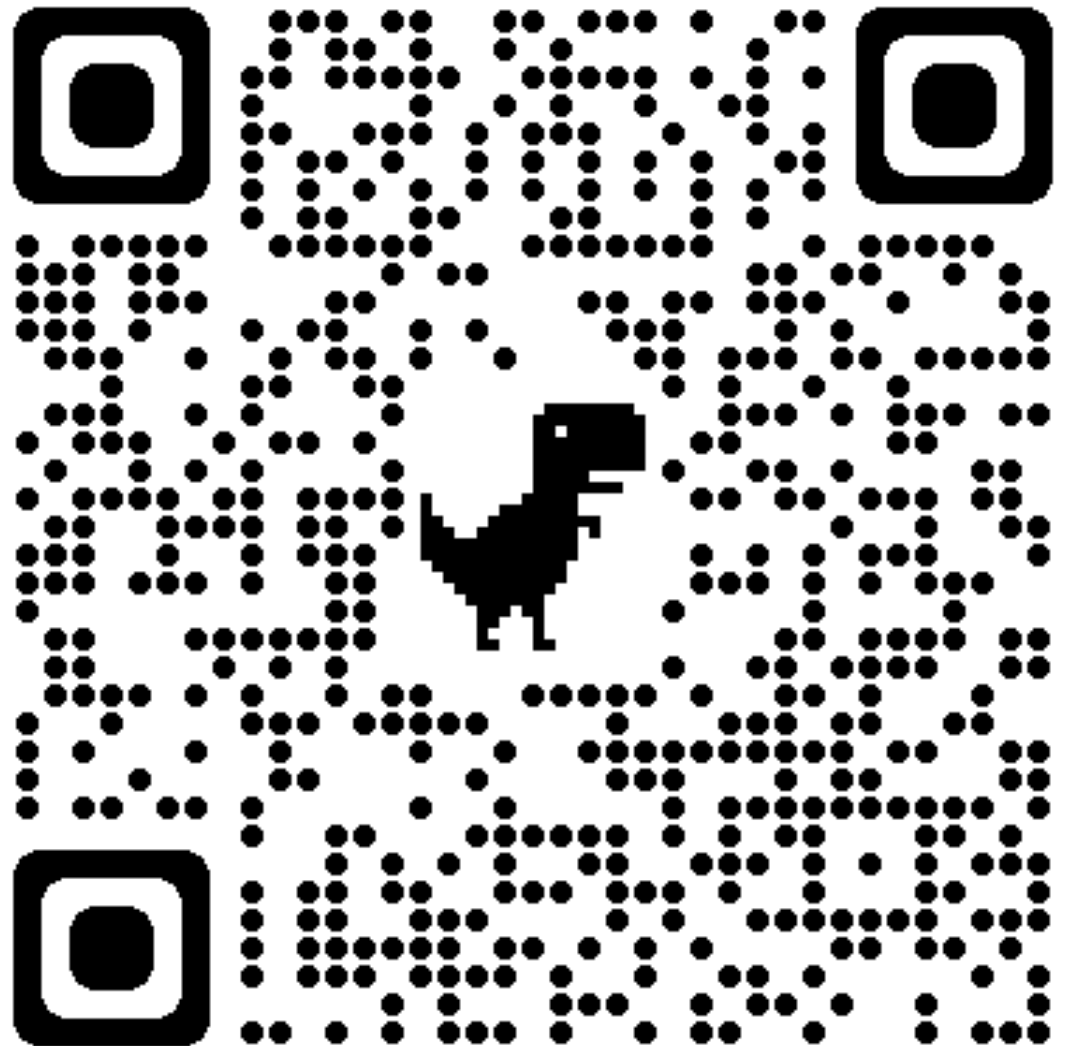
Metabolism: Designed to push the mean information flow towards zero.
 Output (response) independent of Input (perturbation)

Control

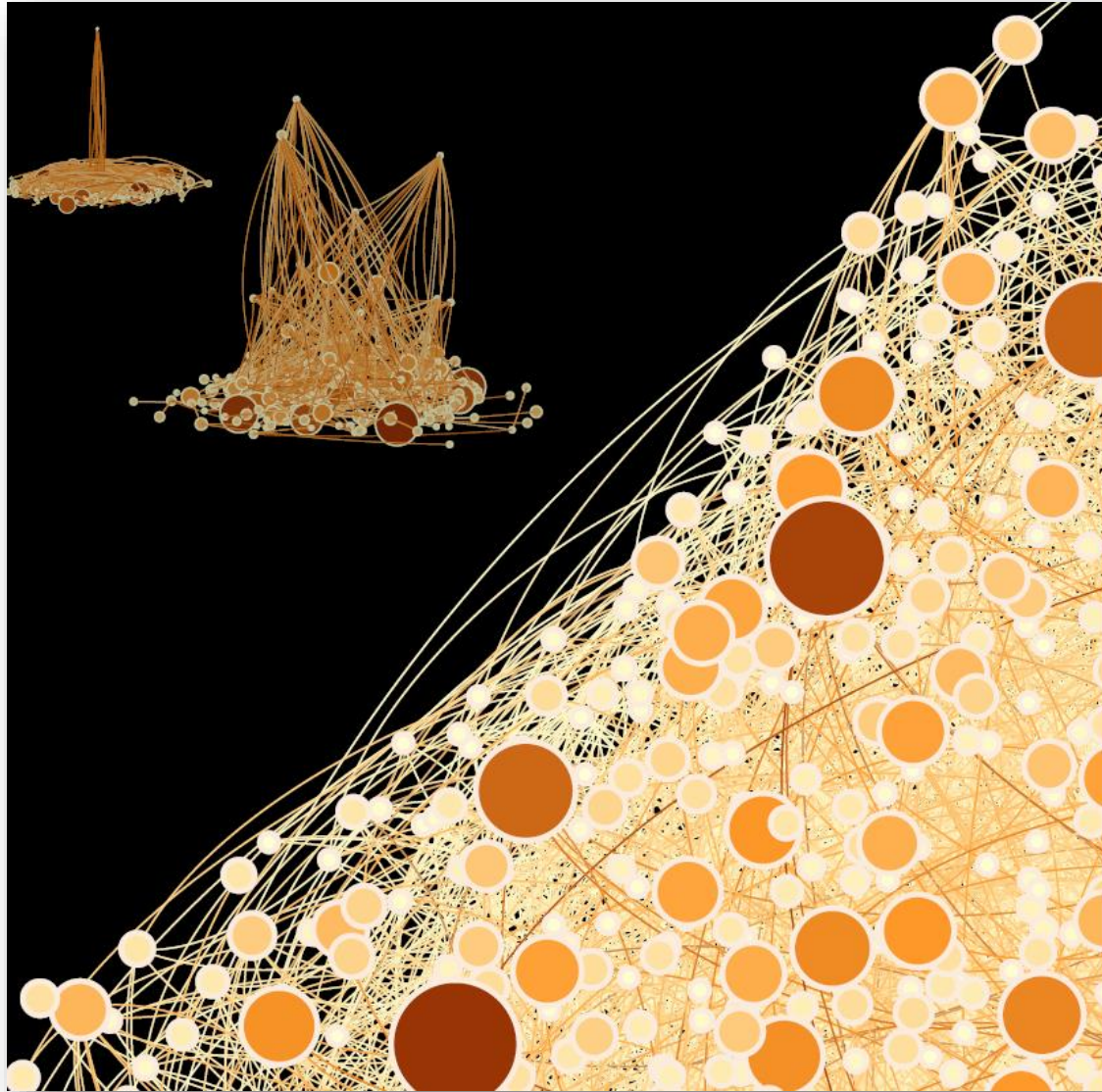


Inaugural meeting on network dynamics & networks of networks

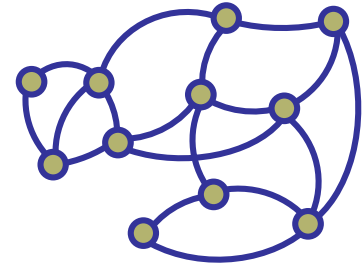
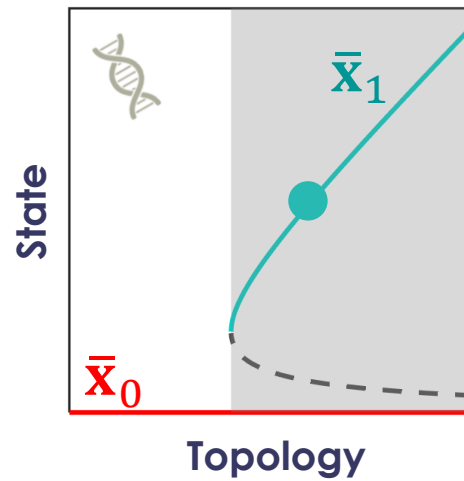
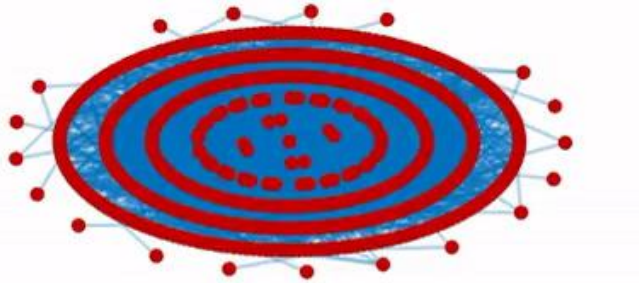
January 29 – February 1, 2024 | Yehuda Hotel, Jerusalem



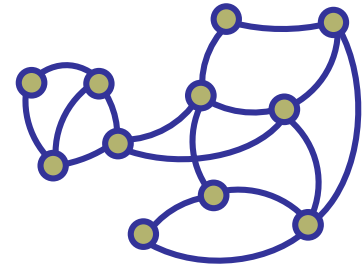
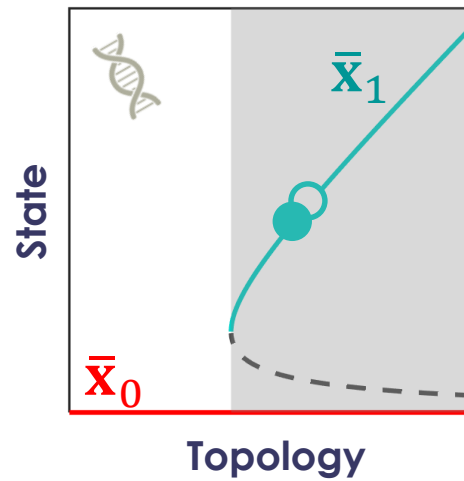
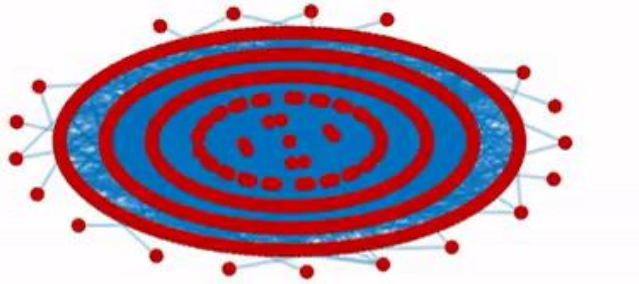
Control



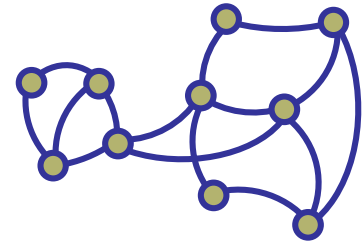
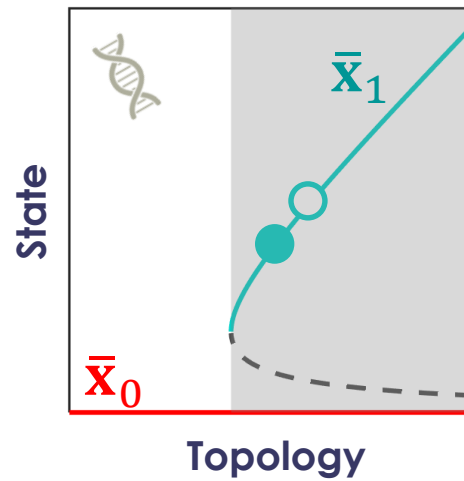
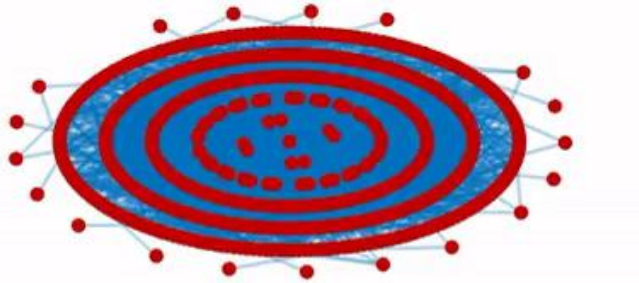
Dynamic transitions



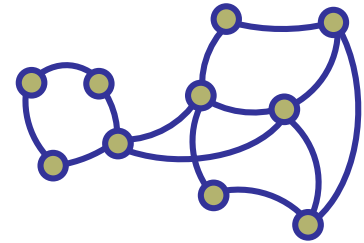
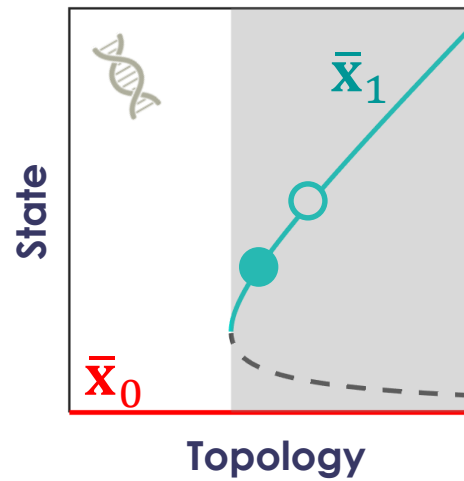
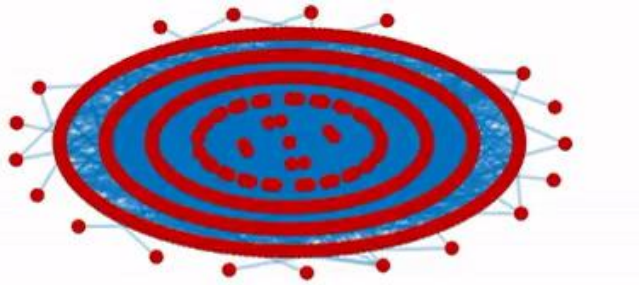
Dynamic transitions



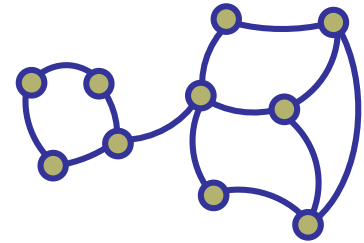
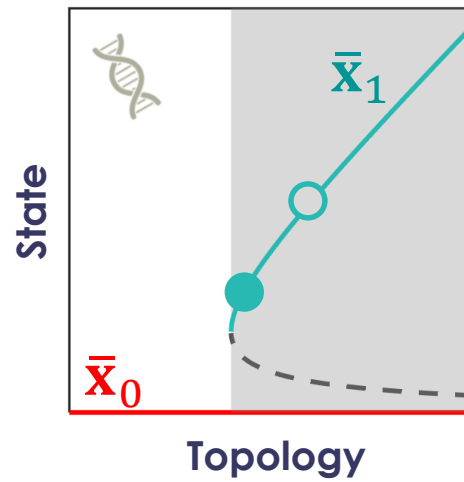
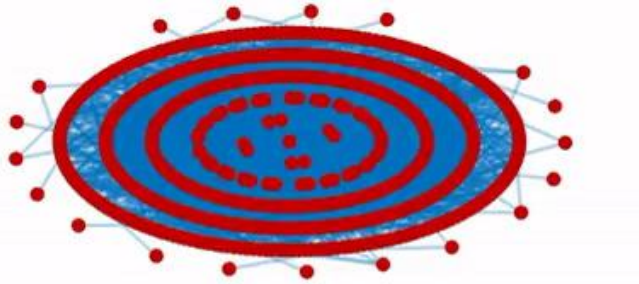
Dynamic transitions



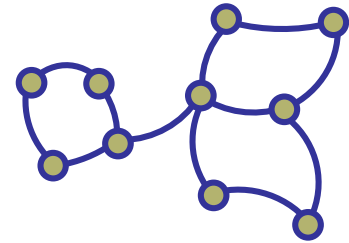
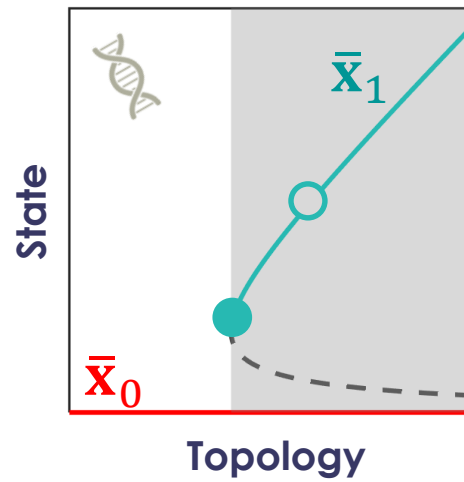
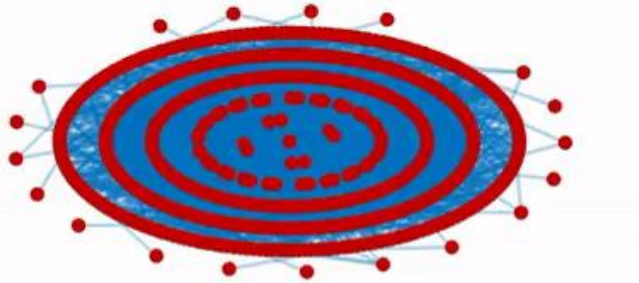
Dynamic transitions



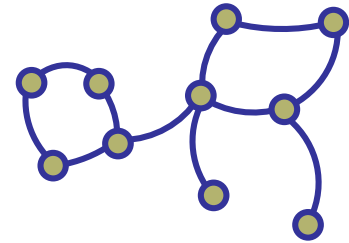
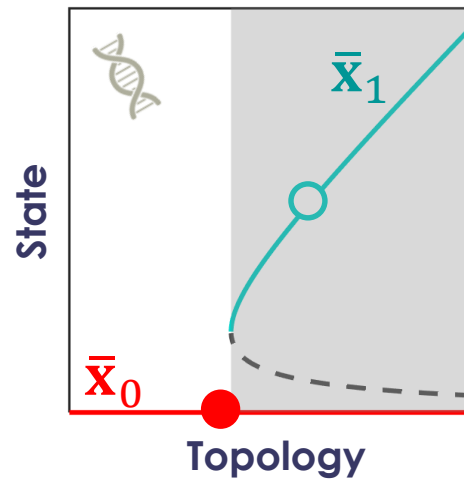
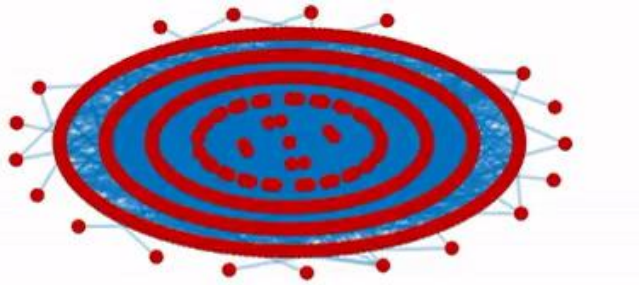
Dynamic transitions



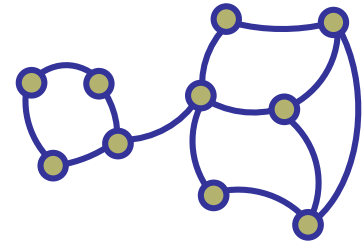
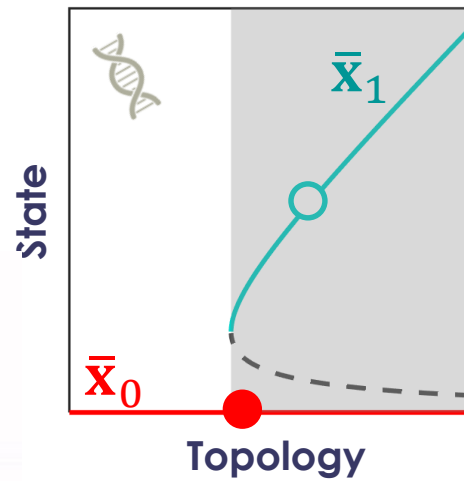
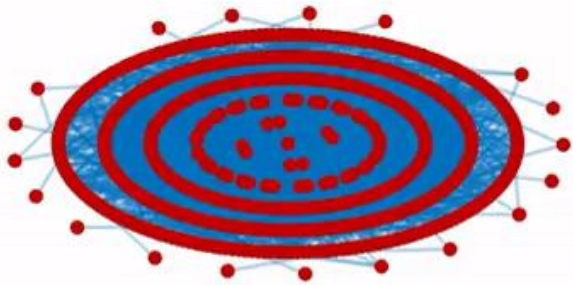
Dynamic transitions



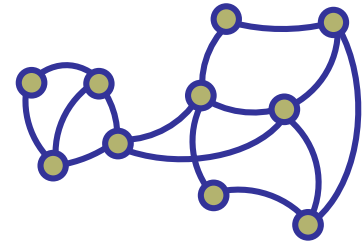
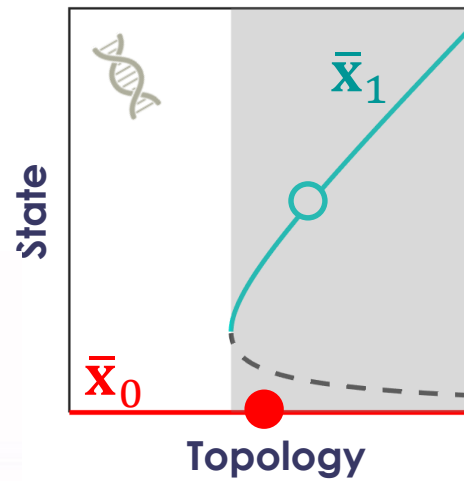
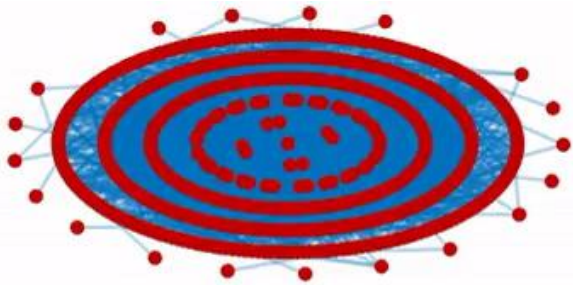
Dynamic transitions



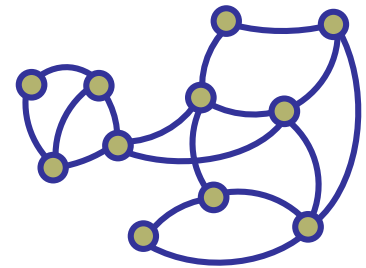
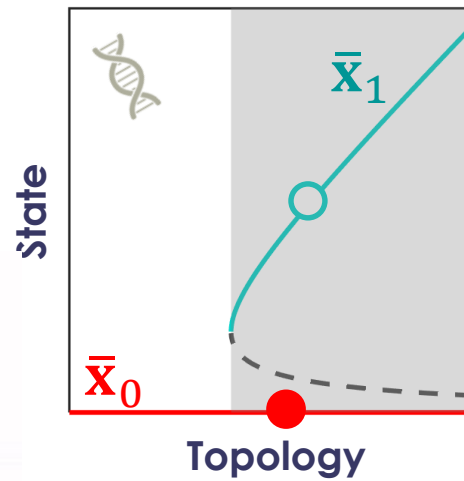
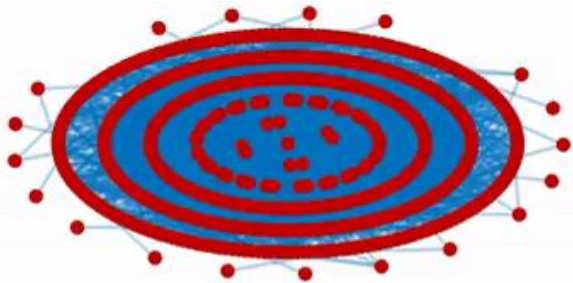
Dynamic transitions - irreversible



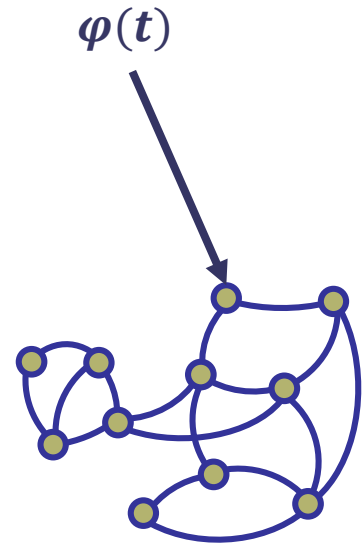
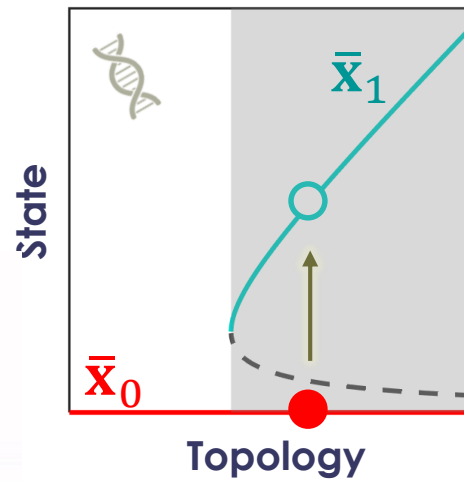
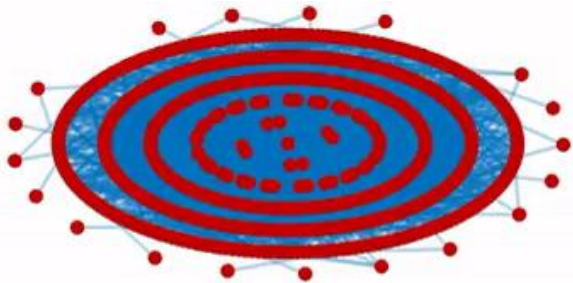
Dynamic transitions - irreversible



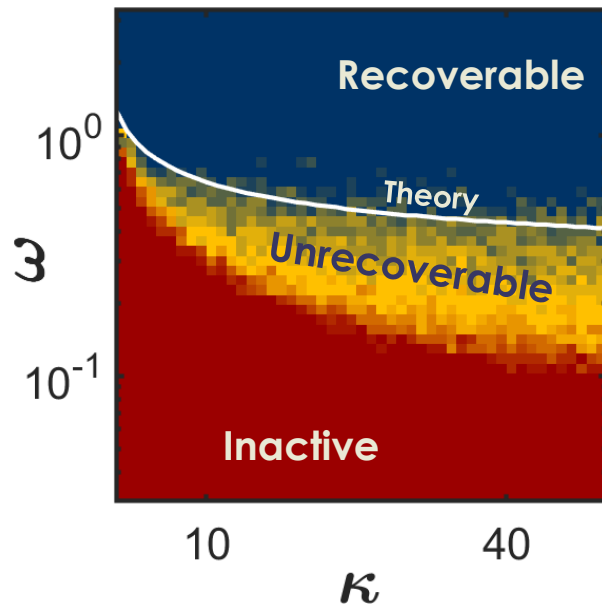
Dynamic transitions - irreversible



Reigniting the network activity

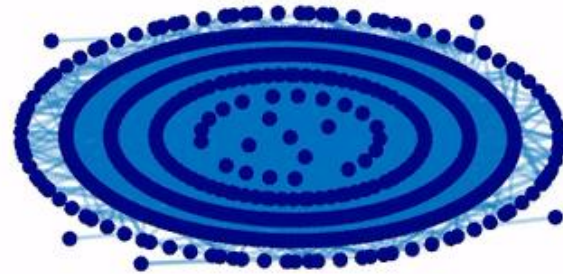
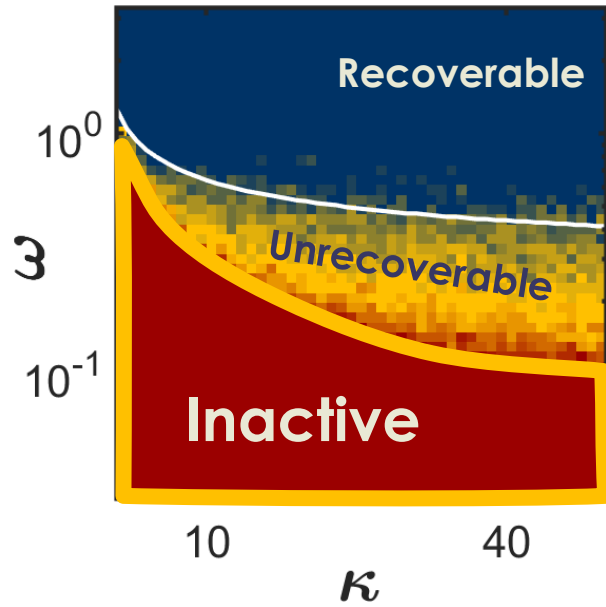


The recoverable phase

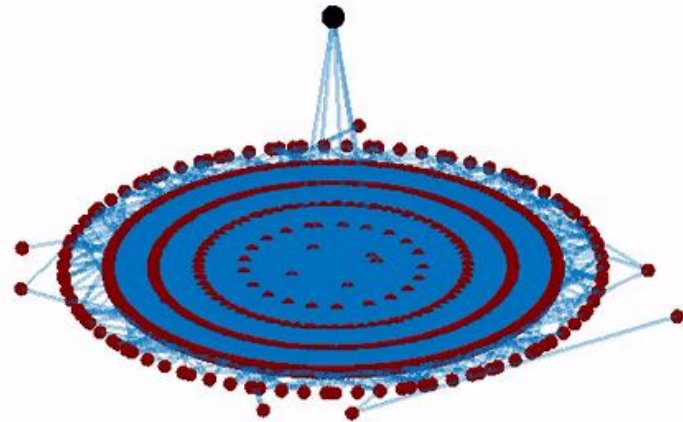
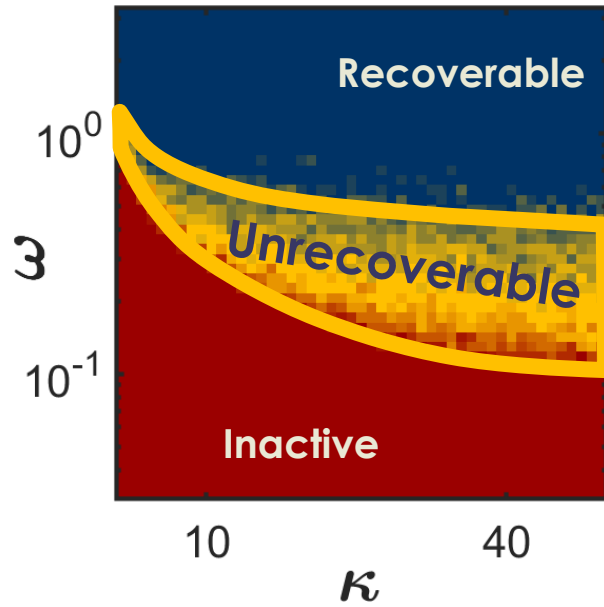


Can you reignite a failed system by controlling just one node?

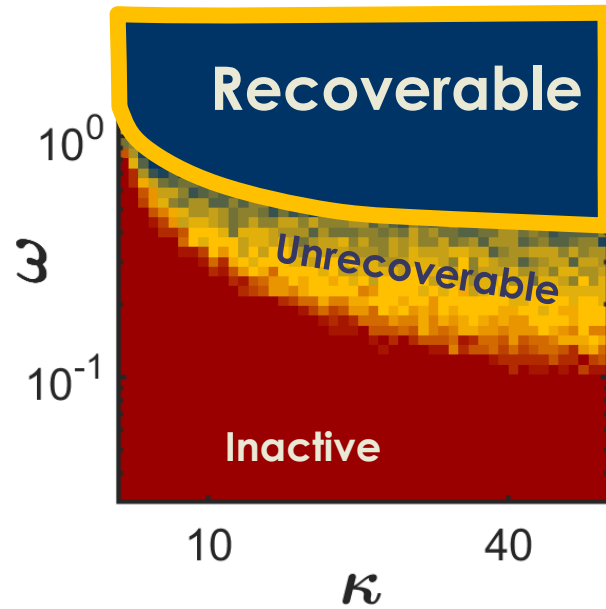
The recoverable phase



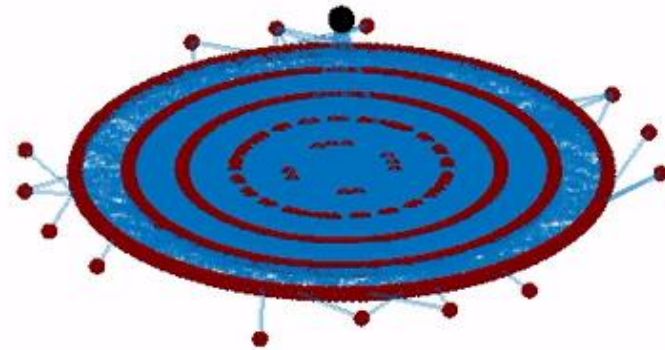
The recoverable phase



The recoverable phase



Single node reigniting –
reviving the failed system by
activating one node



Theory of network dynamics was brought to you by



Guy Berger



Dr. Chittaranjan Hens



Dr. Chandrakala Meena



Dr. Aradhana Singh



Uzi Harush



Dr. Suman Acharya



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