
Controlling Nonlinear Dynamics in Complex Networks

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Preliminary Remarks on Linear Dynamics

Theory

$$\frac{dx(t)}{dt} = Ax(t) + Bu(t)$$

$A_{ii} = 0$ Y. Y. Liu, J. J. Slotine, and A.-L. Barabási, Nature **473**, 167 (2011).

number of control inputs is determined mainly by ... degree distribution

$A_{ii} \neq 0$ N. J. Cowan, E. J. Chastain, D. A. Vilhena, J. S. Freudenberg, and C. T. Bergstrom, PLoS ONE **7**, e38398 (2012).

a single control input ... is all that is needed for structural controllability

C.-T. Lin, IEEE Trans. Autom. Control. **19**, 201 (1974).

K. Murota, *Systems Analysis by Graphs and Matroids: Structural Solvability and Controllability* (1st ed., Springer, Berlin, 1987).

Theory

$$\frac{dx(t)}{dt} = Ax(t) + Bu(t)$$

Kalman's controllability matrix:

$$K = [B \ AB \ \dots \ A^{n-1}B]$$

Can the control signal be constructed in practice?

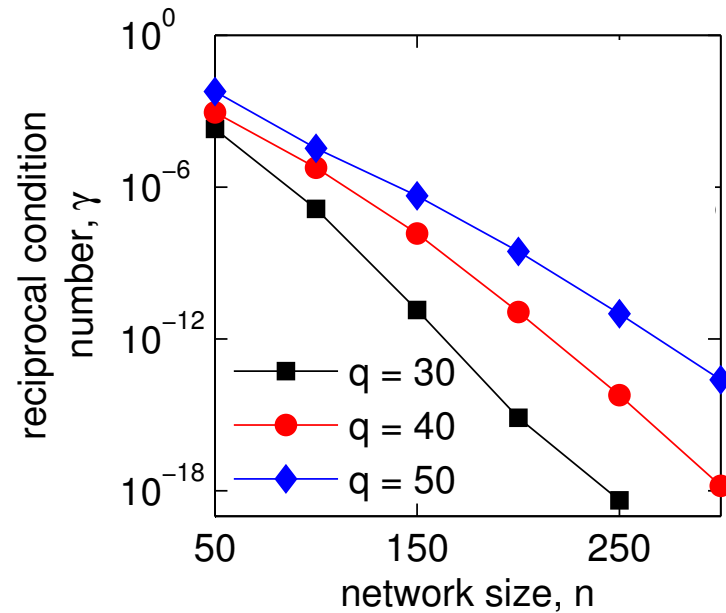
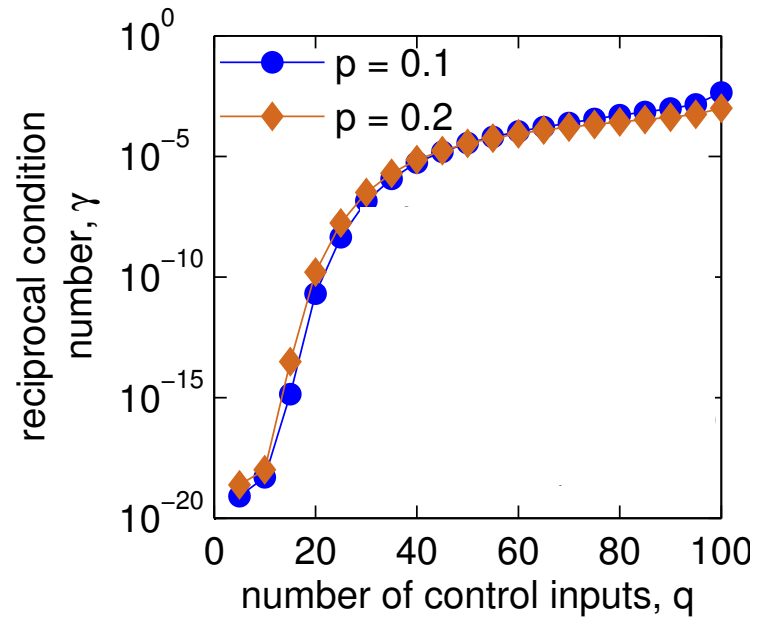
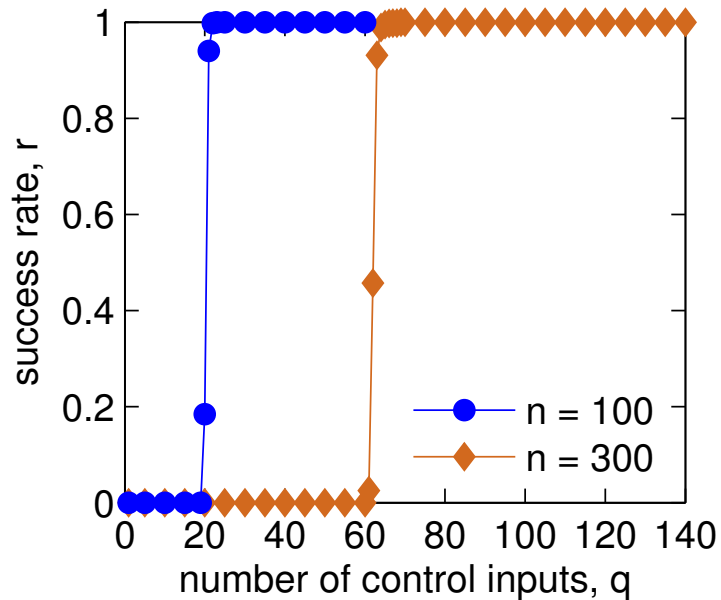
$$u(t) = B^T \Phi^T(t_0, t) W^{-1}(t_0, t_1) [\Phi(t_0, t_1) x^{(1)} - x^{(0)}]$$

$$x(t) = \Phi(t, t_0) [x^{(0)} + W(t_0, t) W^{-1}(t_0, t_1) (\Phi(t_0, t_1) x^{(1)} - x^{(0)})]$$

Controllability Gramian:

$$W(t_0, t_1) = \int_{t_0}^{t_1} \Phi(t_0, t) B B^T \Phi^T(t_0, t) dt, \quad \Phi(t', t) = e^{(t'-t)A}$$

Numerics



Theory for numerics

Perturbation analysis:

$$\|\tilde{x}^{(1)} - x^{(1)}\| \lesssim D \|W(\tilde{W}^{-1} - W^{-1})\|$$

Critical reciprocal condition number:

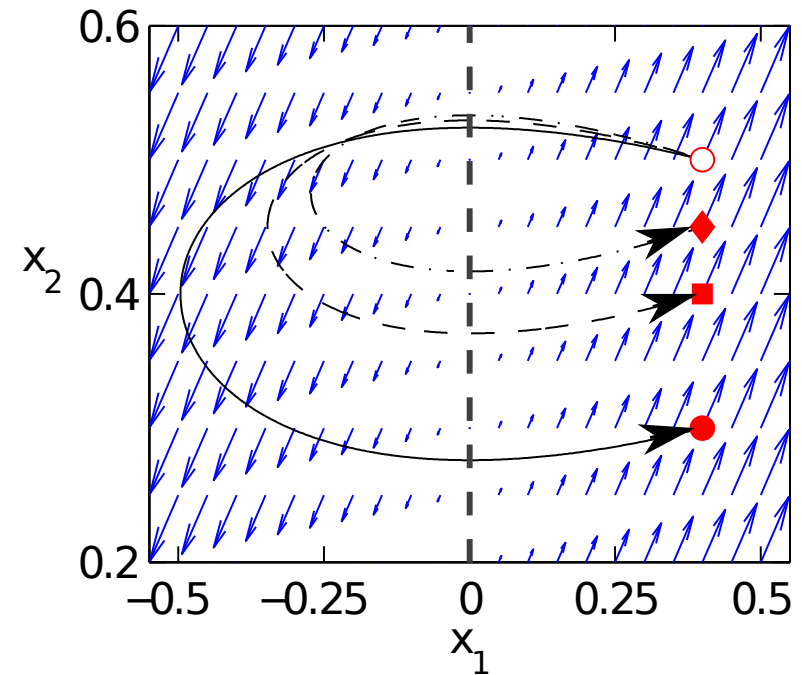
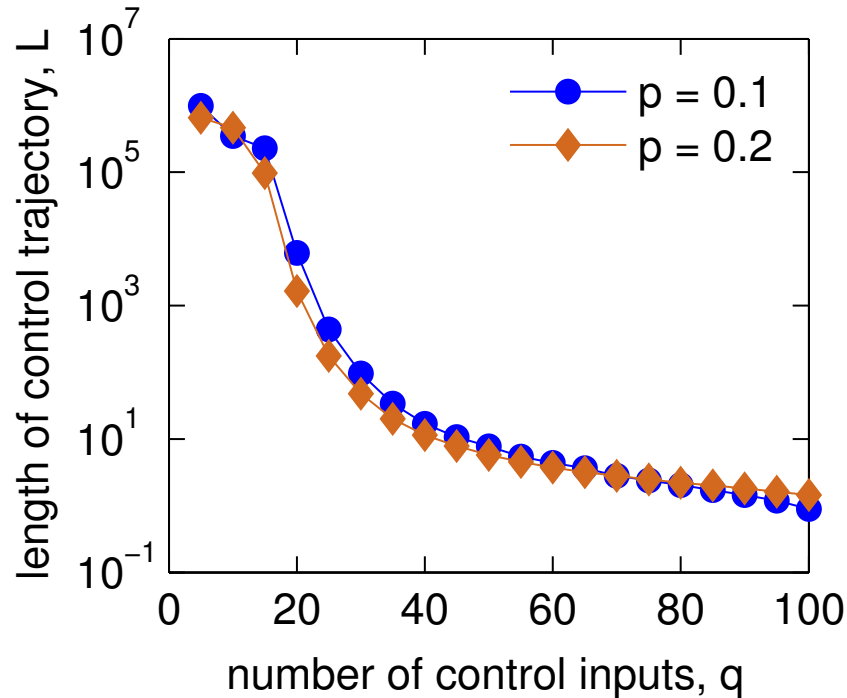
$$\gamma_c \approx D' \frac{\epsilon}{\eta}$$

This leads to the notion of numerical rank condition

Control trajectories

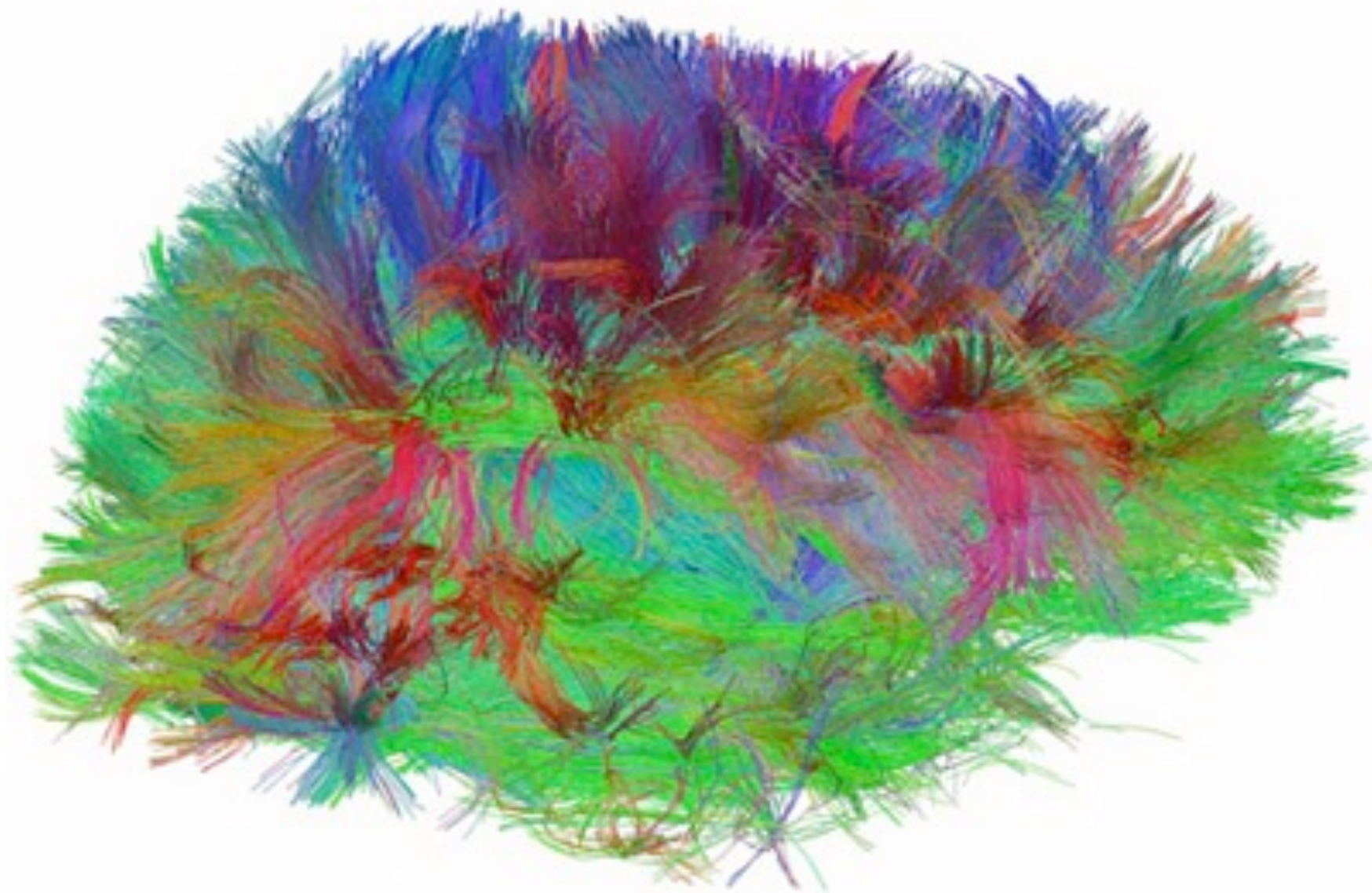
$$\dot{x}_1 = x_1 + u_1(t)$$

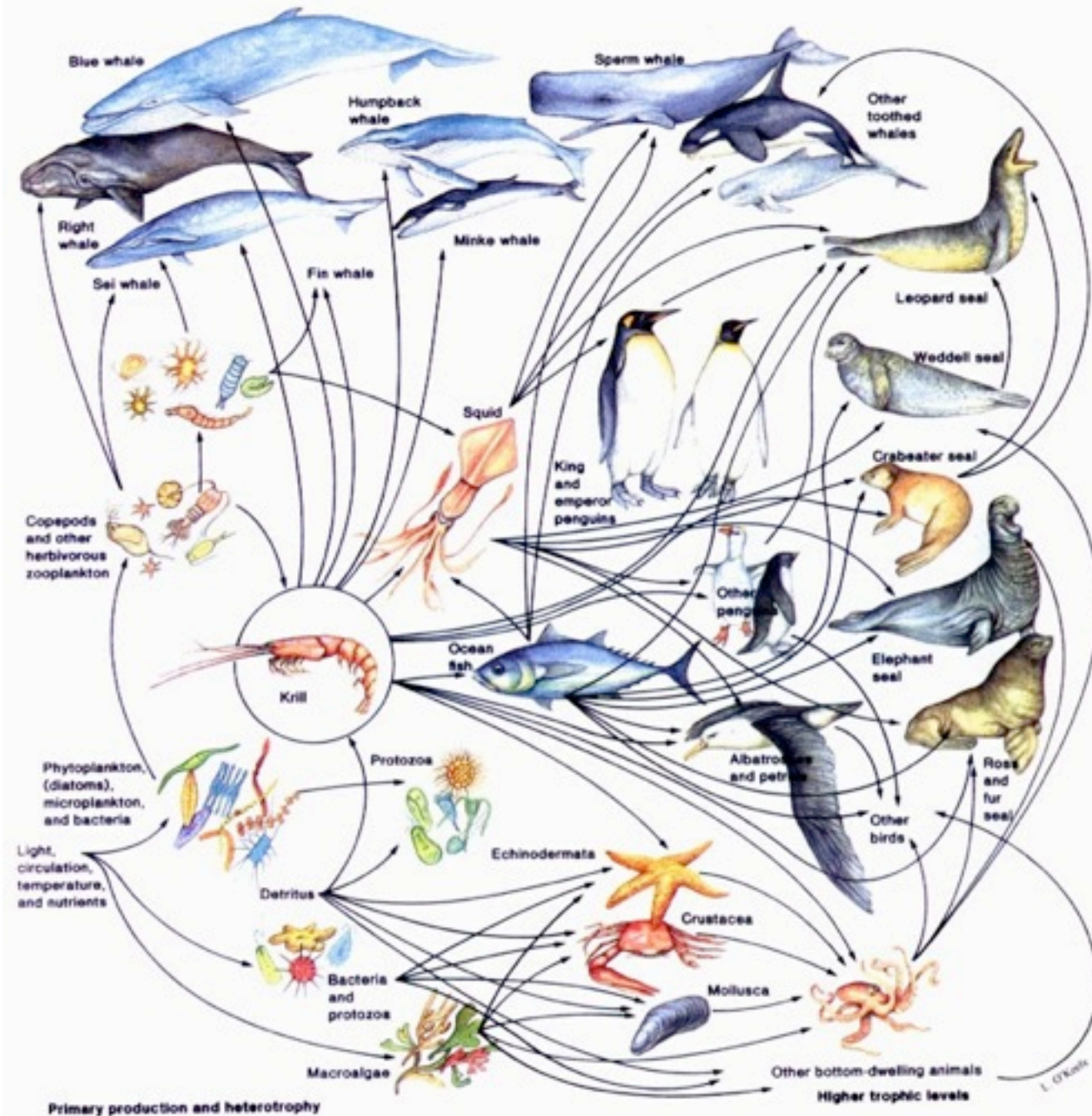
$$\dot{x}_2 = x_1$$

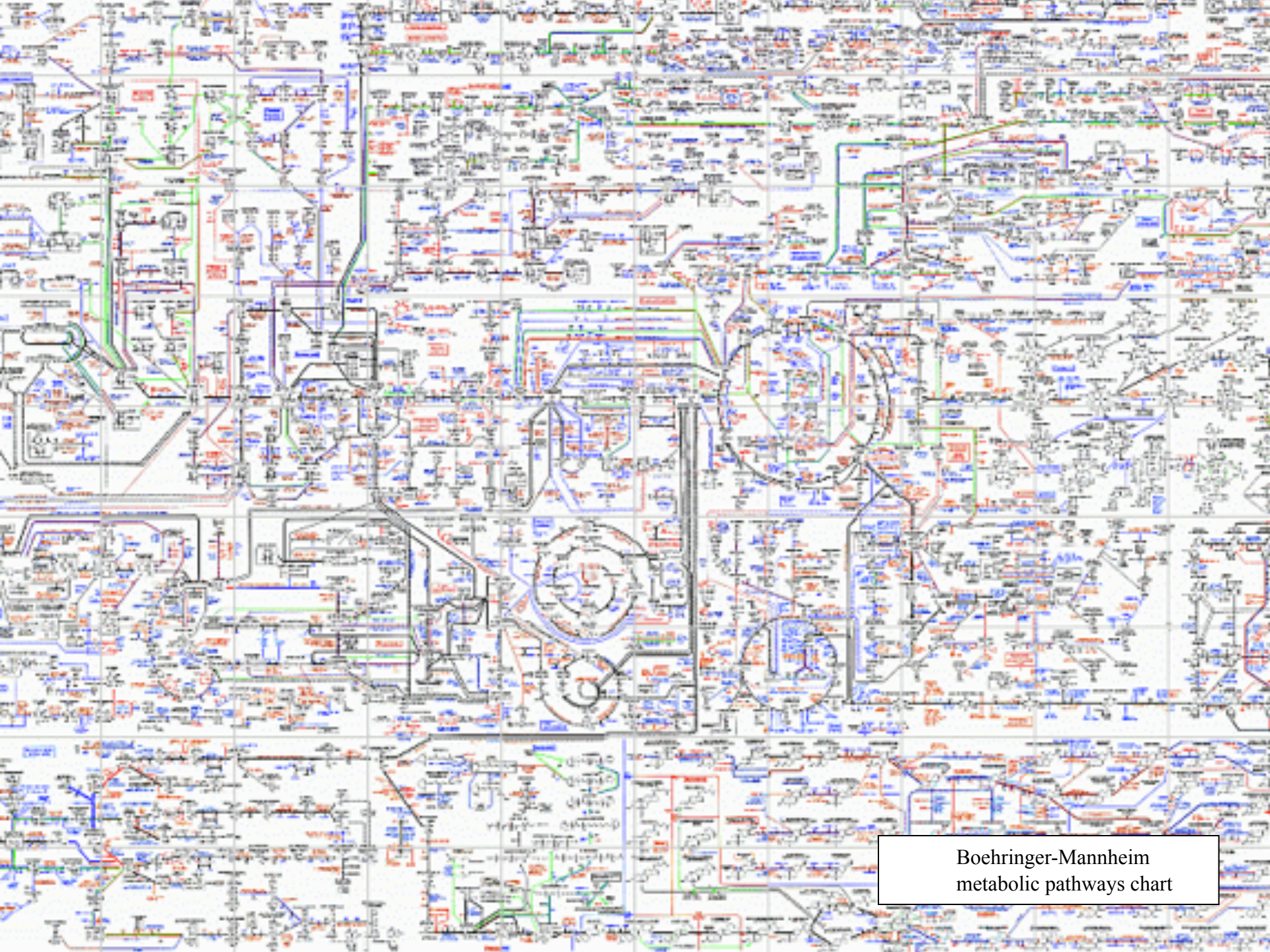


Control trajectories are nonlocal!

Now Turning to Real (Nonlinear) Systems







Boehringer-Mannheim
metabolic pathways chart



Common properties

- (i) the dynamics is nonlinear
 - (ii) the system has multiple stable states (or attractors)
 - (iii) the system is described by a large number of dynamical variables
 - (iv) there are constraints on the physically feasible control interventions
 - (v) there might be noise and parameter uncertainty
 - (vi) decentralized (hence suboptimal) response to perturbations
-

Dynamical equations

Power grids:

$$M_i \frac{d\omega_i}{dt} = P_{mi} - P_{ei}, \quad \frac{d\delta_i}{dt} = \omega_i$$

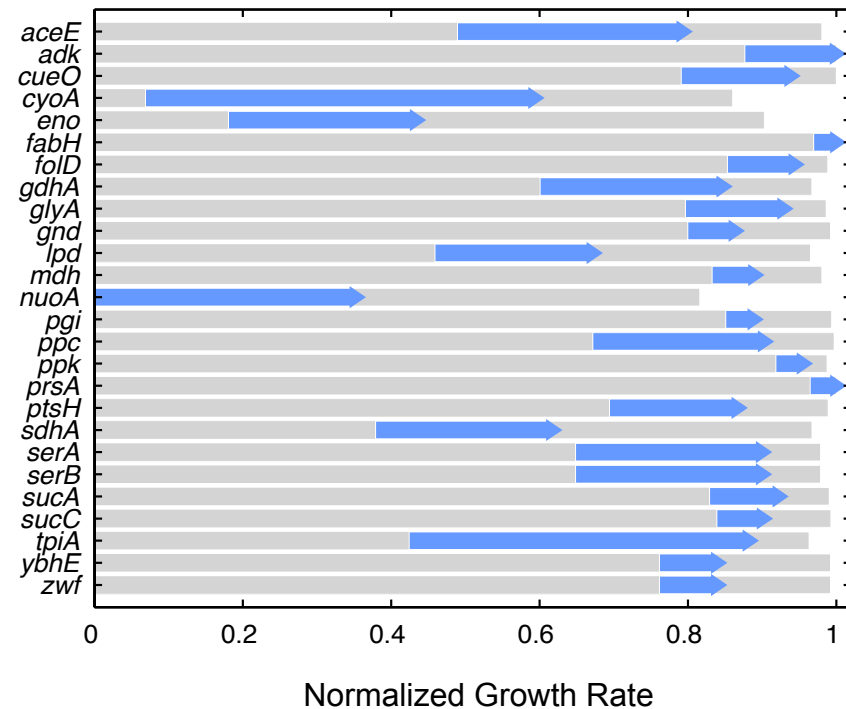
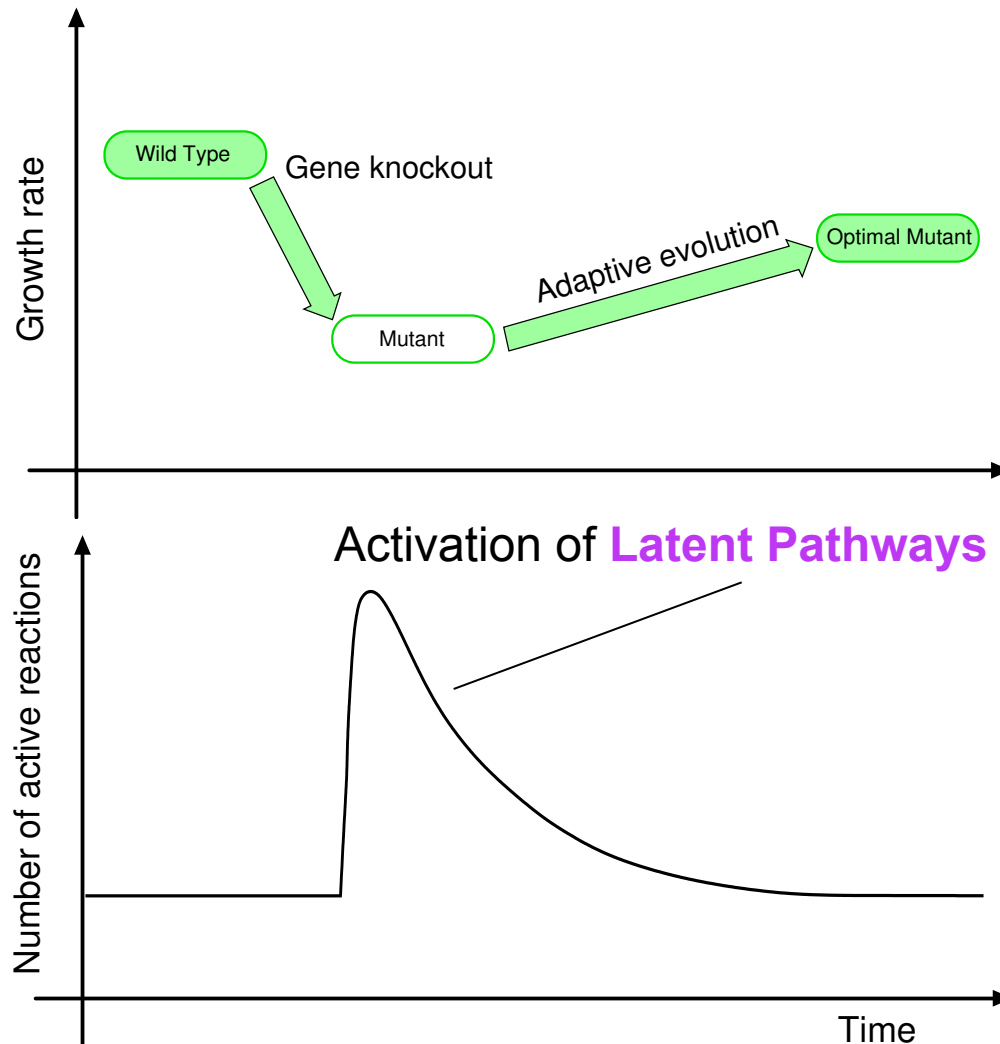
Metabolism:

$$\frac{dX_i}{dt} = \sum_j S_{ij} \nu_j, \quad \nu_j = k_j \prod_i X_i^{\kappa_{ij}}$$

Food webs:

$$\frac{dX_i}{dt} = X_i (b_i + \sum_j a_{ij} X_j)$$

Synthetic rescues in metabolic networks



webs



General Problem

$$\frac{d\mathbf{x}}{dt} = \mathbf{F}(\mathbf{x}; \boldsymbol{\beta})$$

state: $\mathbf{x} \in D \subset \mathbb{R}^m$

parameters: $\boldsymbol{\beta} \in S \subset \mathbb{R}^M$

vector field: $\mathbf{F} : D \times S \rightarrow \mathbb{R}^m$

attraction basin: $\Omega_{\boldsymbol{\beta}}(A) \subset D$

$(\mathbf{x}_0 \in \Omega_{\boldsymbol{\beta}}(A) \text{ iff } \phi_{\boldsymbol{\beta}}(t, \mathbf{x}_0) \rightarrow A \text{ as } t \rightarrow +\infty)$

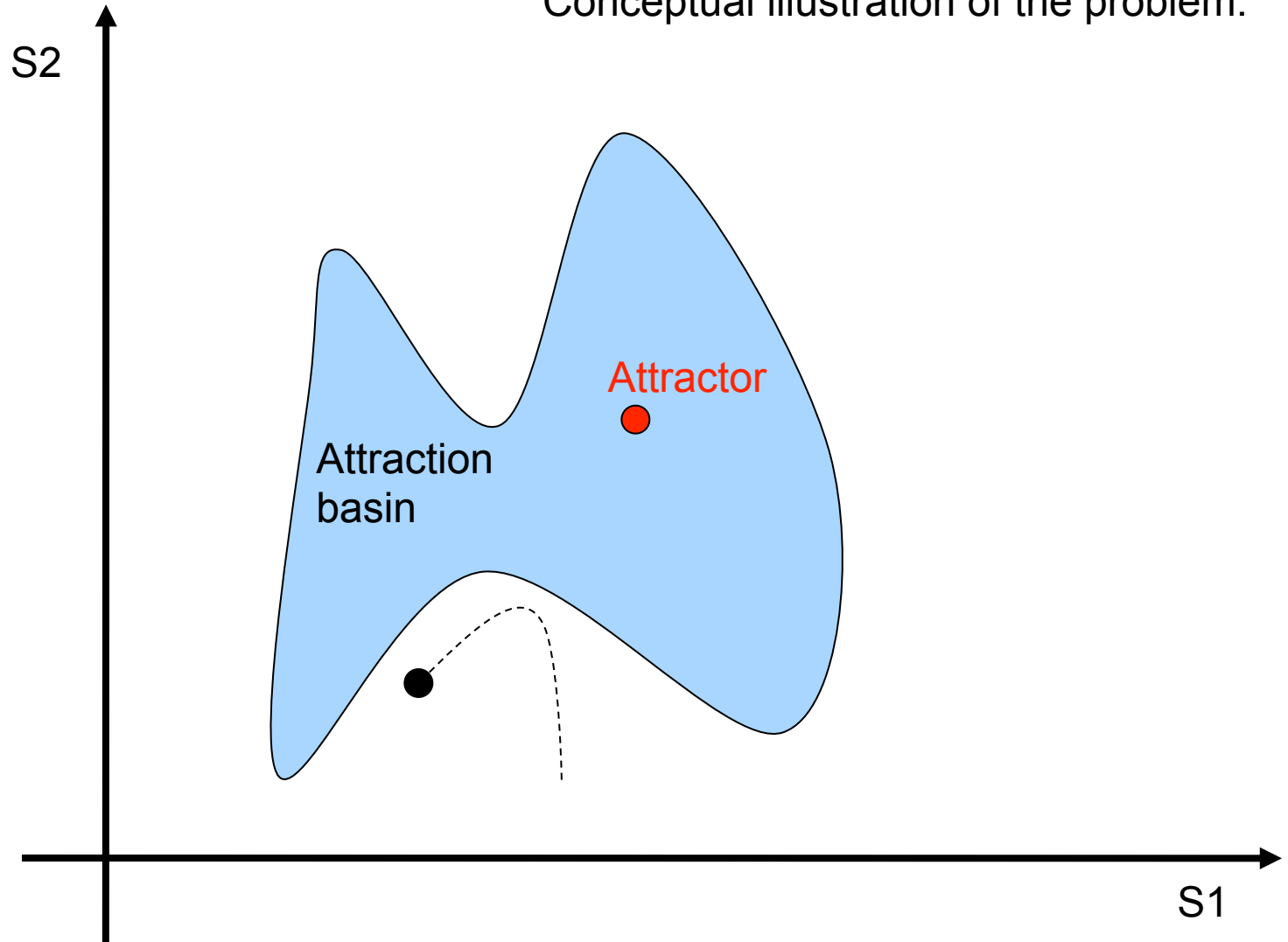
Given $\mathbf{x}_0 \notin \Omega_{\boldsymbol{\beta}}(A)$, find $\Delta \mathbf{x}_0^A$ such that

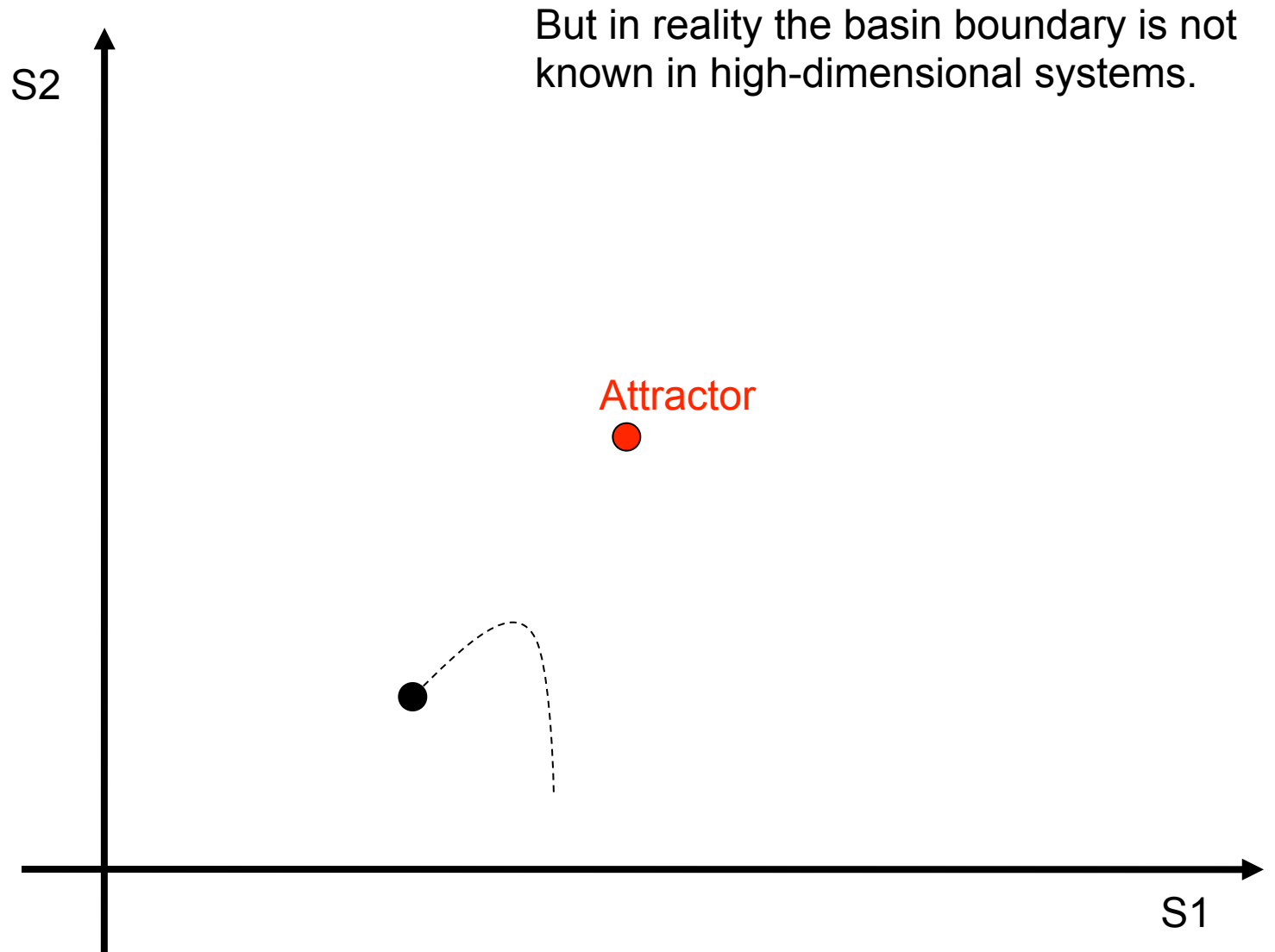
$$\mathbf{x}'_0 \equiv \mathbf{x}_0 + \Delta \mathbf{x}_0^A \in \Omega_{\boldsymbol{\beta}}(A)$$

$$g_i^x(\mathbf{x}'_0, \mathbf{x}_0) \leq 0, \quad i = 1, \dots, p$$

$$h_j^x(\mathbf{x}'_0, \mathbf{x}_0) = 0, \quad j = 1, \dots, q, \quad \text{where } g_i^x, h_j^x : D \times D \rightarrow \mathbb{R}$$

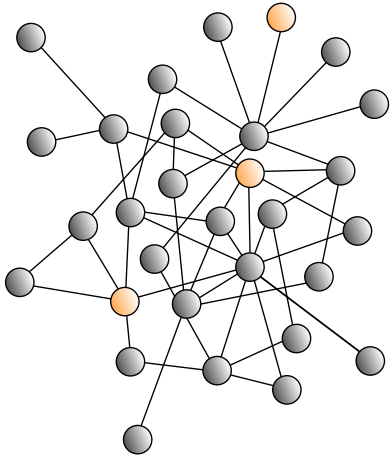
Conceptual illustration of the problem.



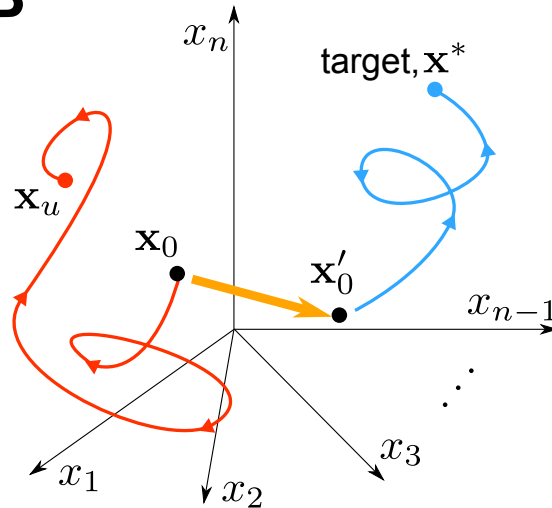


Identifying Control Interventions

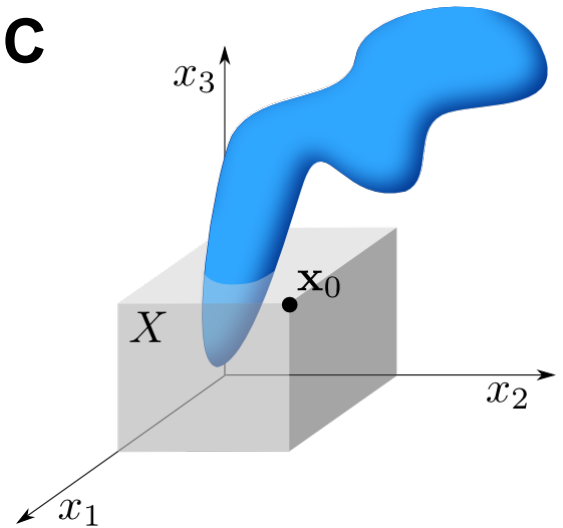
A



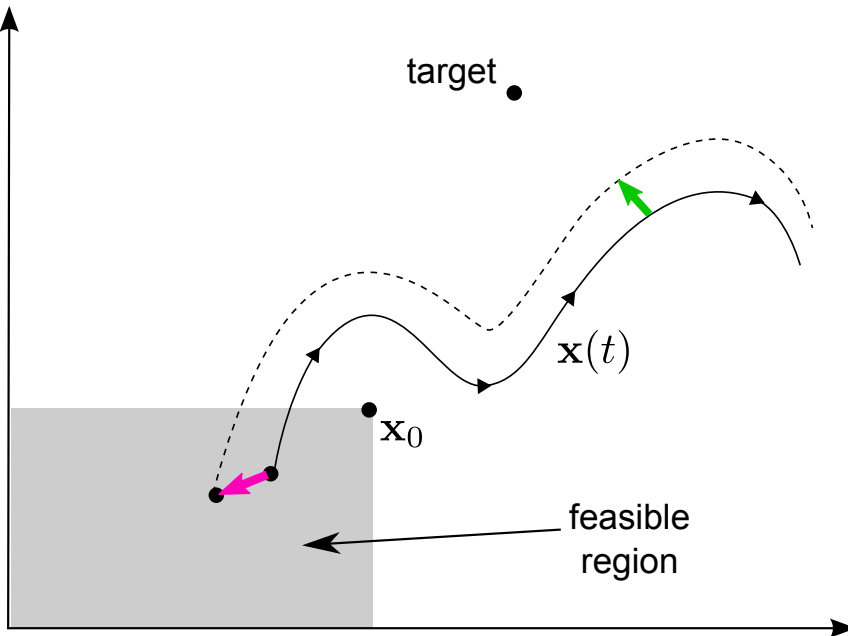
B



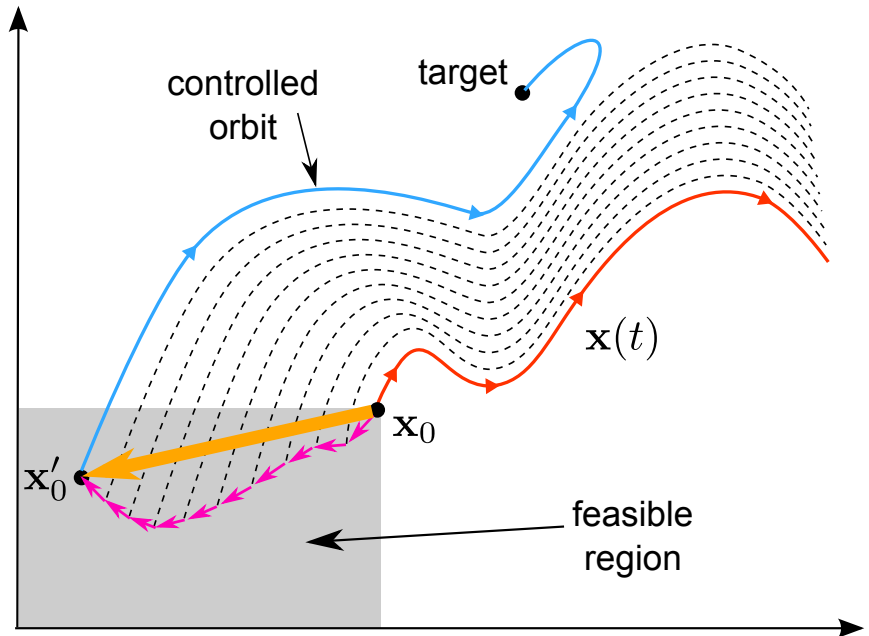
C



D



E



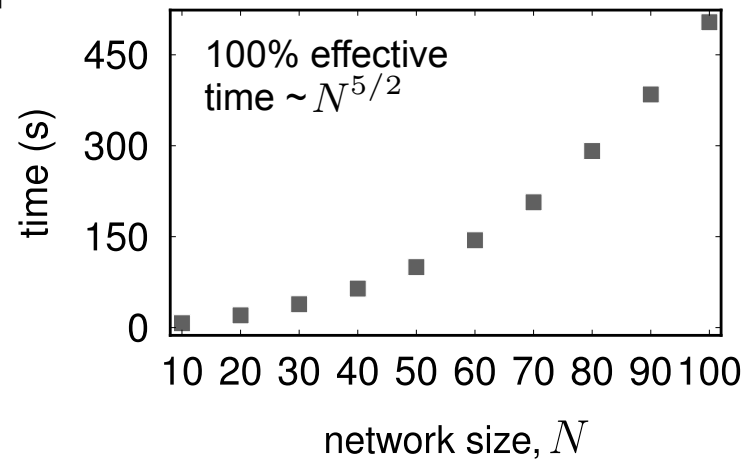
Effectiveness and Efficiency

Units

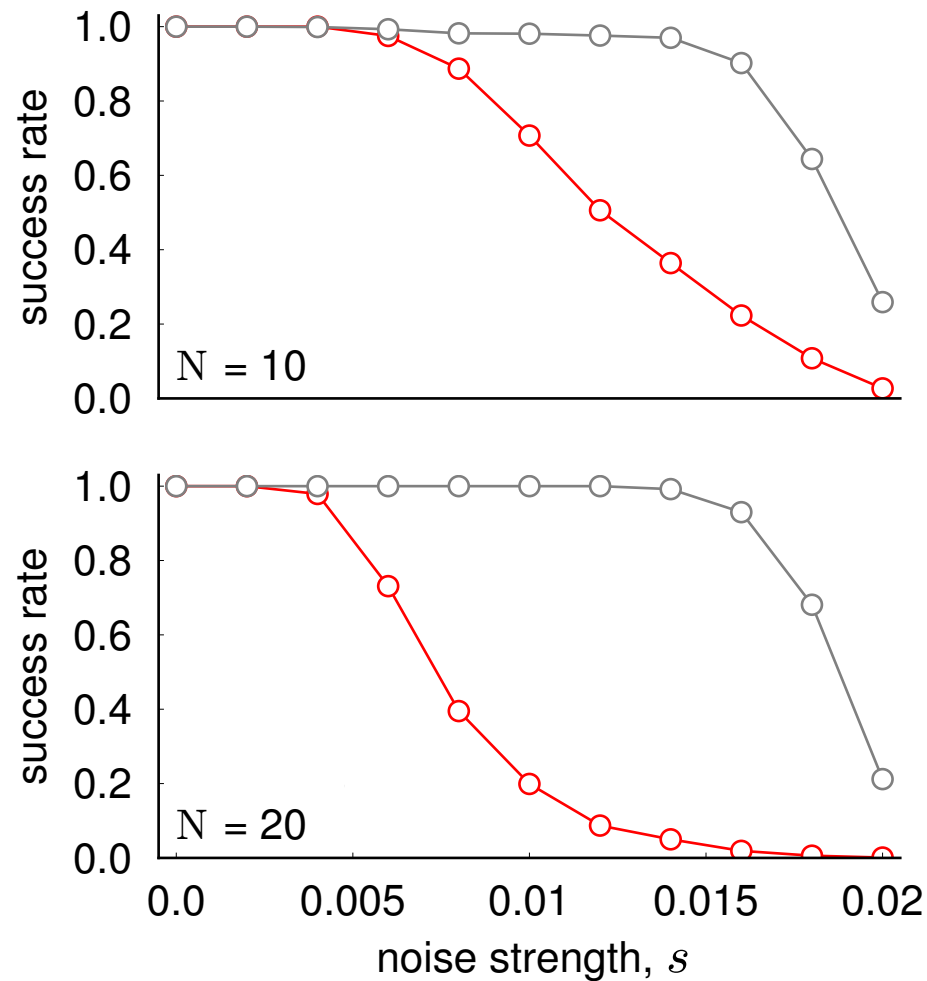
$$\frac{dx_1}{dt} = a_1 \frac{x_1^4}{x_1^4 + S^4} + b_1 \frac{S^4}{x_2^4 + S^4} - k_1 x_1 + f_1$$
$$\frac{dx_2}{dt} = a_2 \frac{x_2^4}{x_2^4 + S^4} + b_2 \frac{S^4}{x_1^4 + S^4} - k_2 x_2 + f_2$$

Network

$$\frac{d\mathbf{x}_i}{dt} = \mathbf{f}(\mathbf{x}_i) + \frac{c}{d_i} \sum_j A_{ij} [\mathbf{x}_j - \mathbf{x}_i]$$

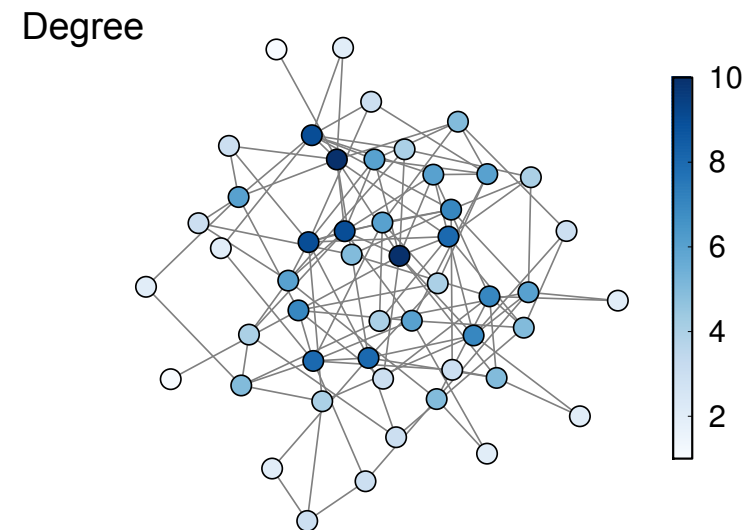
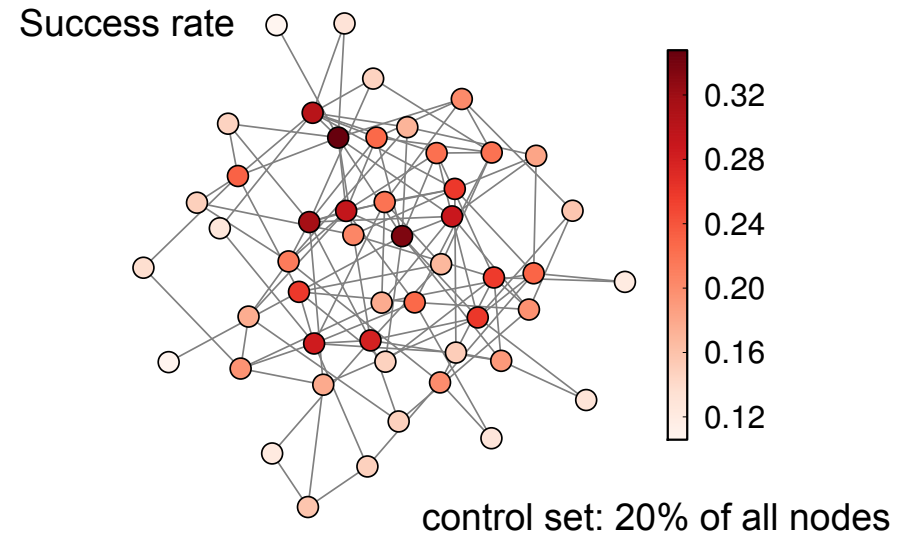
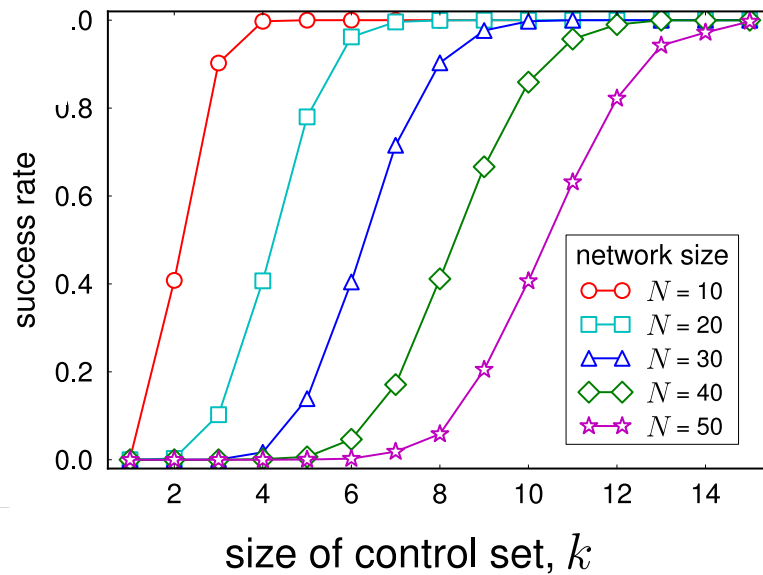


Robust against Stochasticity and Parameter Uncertainty



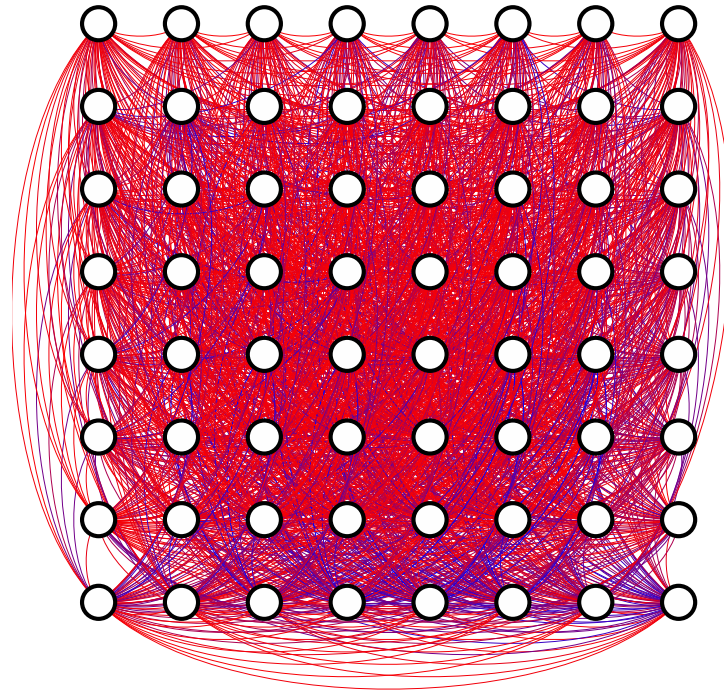
Small Number of Accessible Nodes

Randomly selected nodes

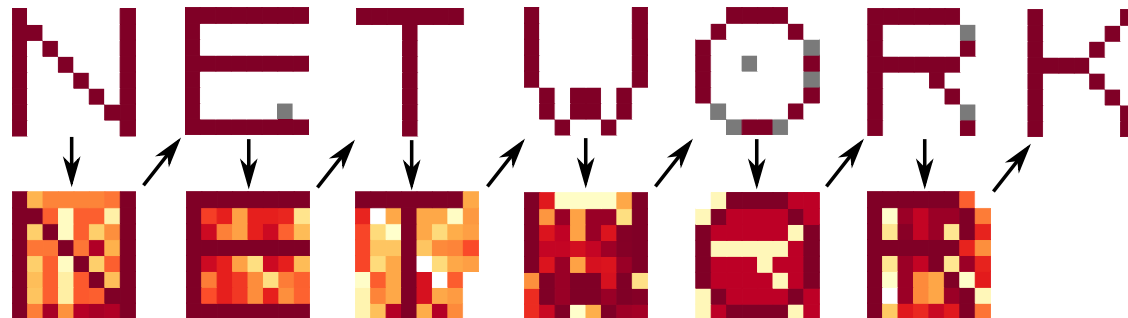


Example

Transitions in associative-memory networks

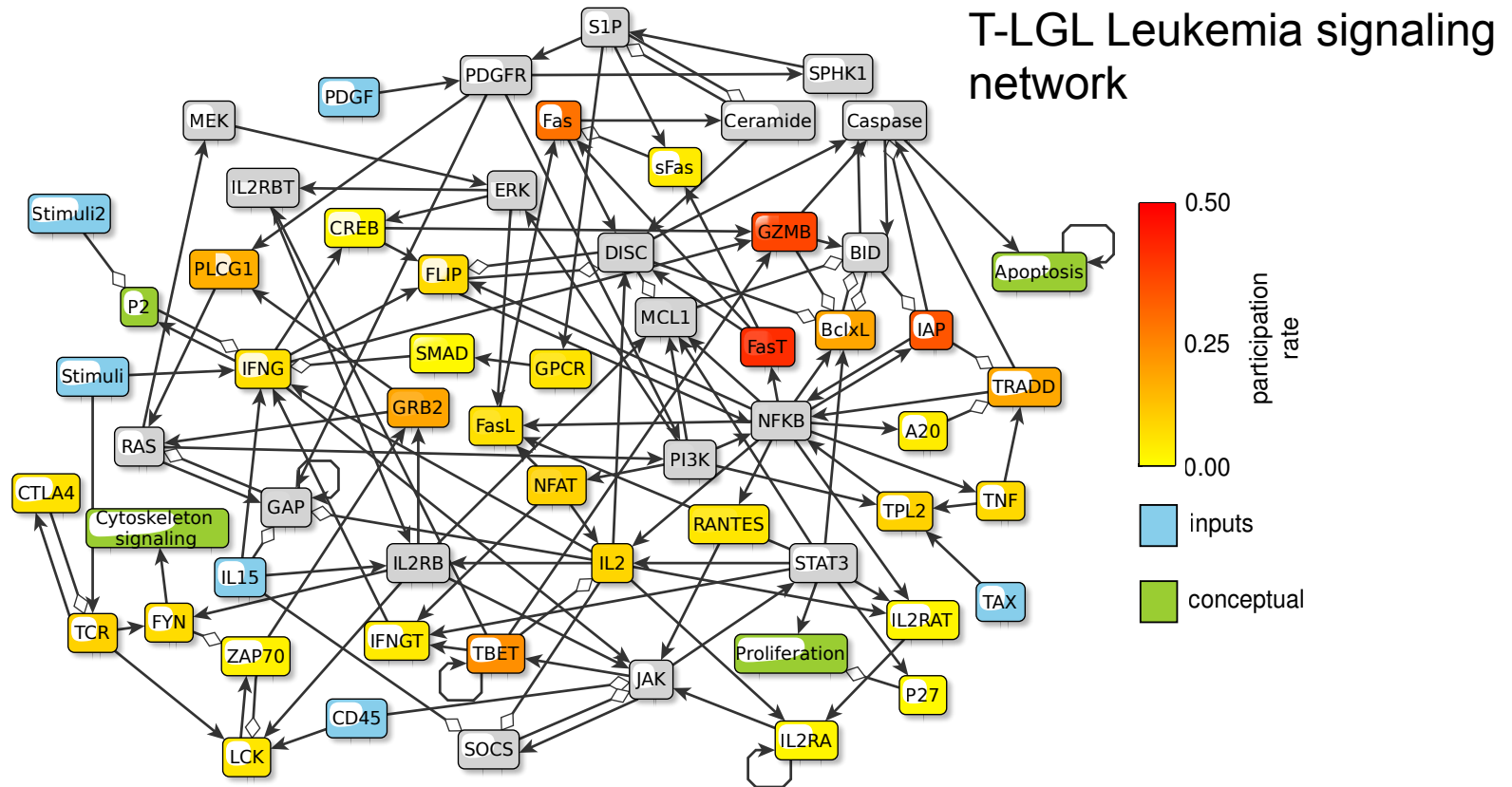


Pattern stored
“NETWORK”



Example

Identification of therapeutic Interventions



Control by Manipulating Stability

Synchronization in power grid networks

Starting from the swing equation

$$\frac{2H_i}{\omega_R} \frac{d^2 \delta_i}{dt^2} = P_{mi} - P_{ei}$$

we can show that stability of synchronous states is enhanced when

$$\beta_i = (D_i + 1/R_i)/2H_i$$

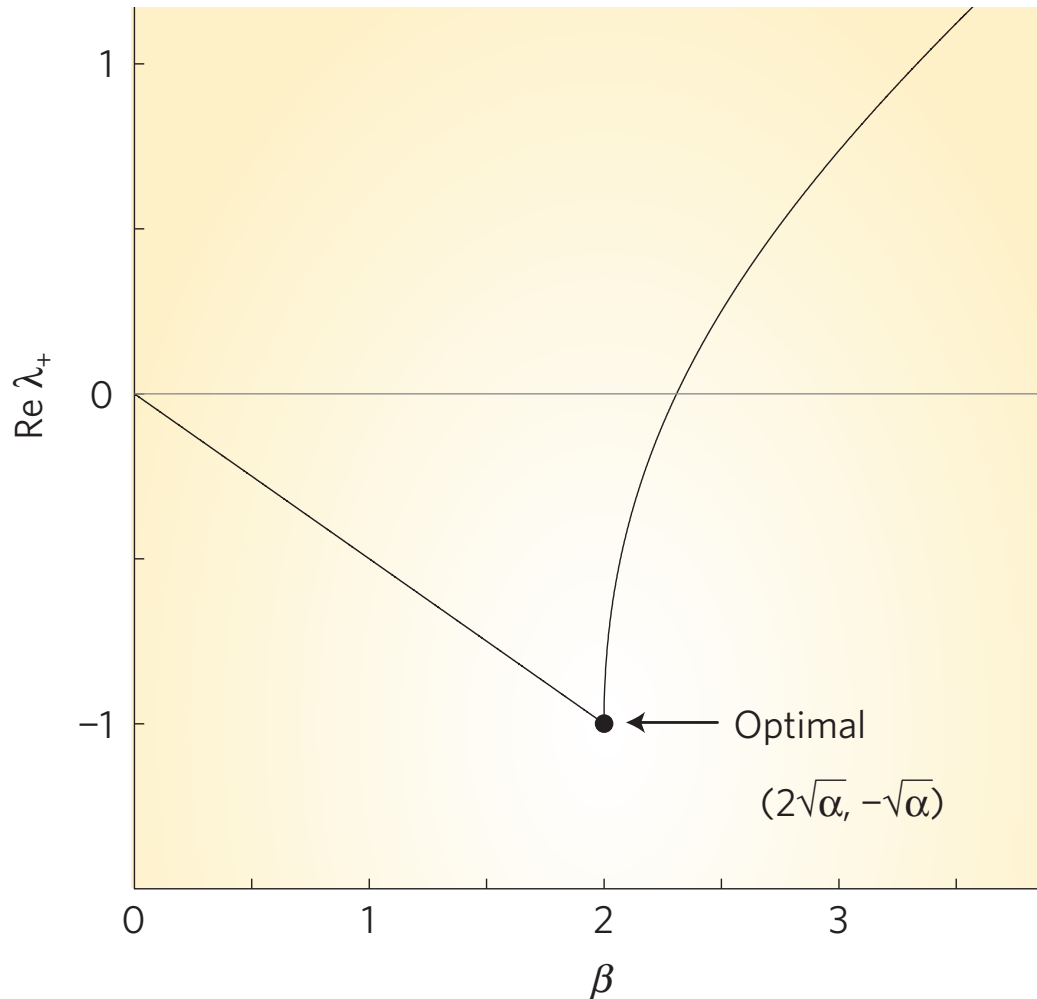
for all generators is equal to $2\sqrt{\alpha_2}$

where α_2 is the smallest nonzero eigenvalue of the “coupling” matrix

$$P_{ij} = \begin{cases} \frac{\omega_R E_i E_j}{2H_i} (G_{ij} \sin \delta_{ij}^* - B_{ij} \cos \delta_{ij}^*), & i \neq j \\ -\sum_{k \neq i} P_{ik}, & i = j \end{cases}$$

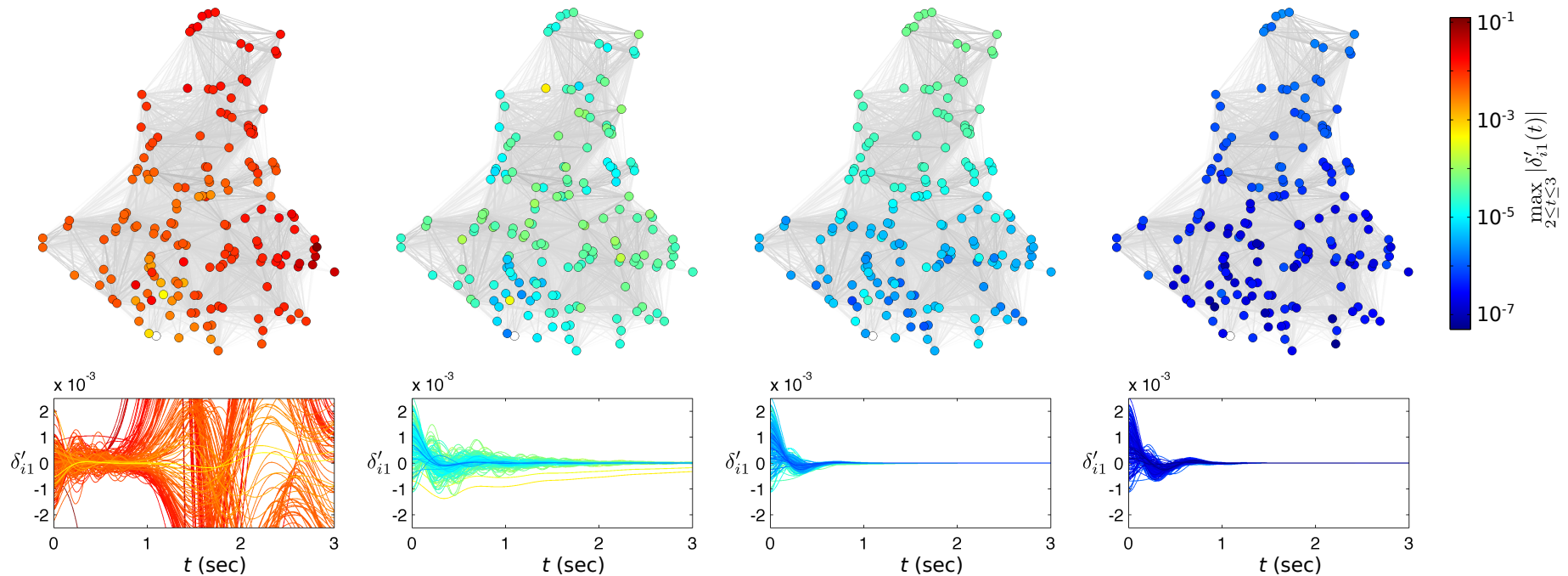
Control by Manipulating Stability

Synchronization in power grid networks



Control by Manipulating Stability

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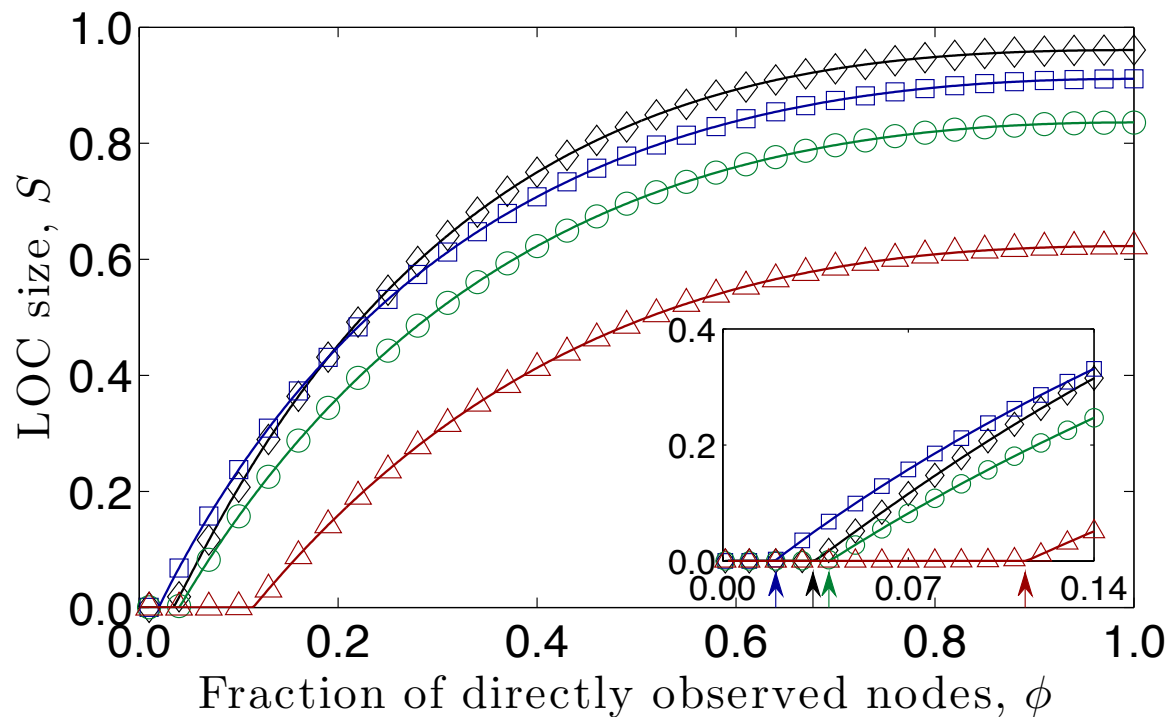


Network Observability Transitions

Example: phasor measurement units in power grids

will allow real time wide area monitoring

new type of percolation transition



Outlook

1. Network response to perturbations can be largely controlled by *eligible* interventions
2. The control interventions can be identified without any *a priori* information about the basins of attraction
3. These interventions can be effective even when the target stable state *cannot* be reached directly
4. It is often the case that globally advantageous interventions are (apparently or in fact) *locally deleterious*
5. We can take advantage of the *nonlinear* (and possibly *noisy*) nature of the dynamics

For more information, please go to the group's
webpage: <http://dyn.phys.northwestern.edu/>
