

# Reconstruction of Complex Dynamical Networks

Wei Pan<sup>1</sup>, Ye Yuan<sup>1,2</sup>, Guy-Bart Stan<sup>1</sup>

1. Centre for Synthetic Biology and Innovation and the Department of Bioengineering, Imperial College London.
2. Department of Engineering, University of Cambridge.

## Background



A group of blind men touch an elephant to learn what it is like. They conclude that the elephant is like a wall, snake, spear, tree, fan or rope, depending upon where they touch. They have a heated debate and the conflict is never resolved.

## Aims

- What is the topology of the network?
- What are the directions and strengths of coupling links between nodes?
- What kind of linear/nonlinear functions govern the dynamics of each node?
- What are the parameters of these functions?

## Model Transformation

$$\dot{x}_i(t) = f_i(x_i(t)) + \sum_{j=1}^n w_{ij} g_{ij}(x_j(t), x_i(t)) + \eta_i(t)$$

$$\dot{x}_{ik}(t) = \sum_{l=1}^{N_{f_i}} w_{f_i}^{(l)}(k) f_i^{(l)}(x_i(t)) + \sum_{j=1}^n \sum_{r=1}^{N_{g_j}} w_{g_j}^{(r)}(k, :) g_j^{(r)}(x_j(t), x_i(t)) + \eta_{ik}(t)$$

$$e_{ik}(t_{T+1}) = F(x_i(t_T)) W_{f_{ik}} + G_i(x(t_T)) W_{g_{ik}} + \xi_{ik}(t_T)$$

$$y_{ik} \triangleq [e_{ik}(t_1), \dots, e_{ik}(t_M)]^T \in \mathbb{R}^{M \times 1}$$

$$\Phi_{ik} \triangleq \begin{bmatrix} F(x_i(t_0)) & G_i(x(t_0)) \\ F(x_i(t_1)) & G_i(x(t_1)) \\ \vdots & \vdots \\ F(x_i(t_{M-1})) & G_i(x(t_{M-1})) \end{bmatrix} \in \mathbb{R}^{M \times N_i}$$

$$x_{ik} \triangleq [W_{f_{ik}}, W_{g_{ik}}] \in \mathbb{R}^{N_i \times 1}$$

$$\Xi_{ik} \triangleq [\xi_{ik}(t_1), \dots, \xi_{ik}(t_{M-1})]^T \in \mathbb{R}^{M \times 1}$$

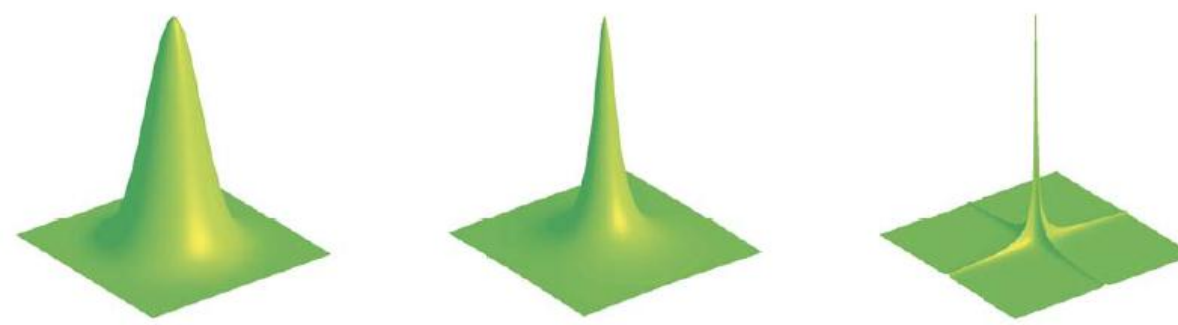
$$y = \Phi x + \Xi$$

$M \times 1$  measurements =  $M \times N$  matrix =  $N \times 1$  sparse signal +  $K$  nonzero entries

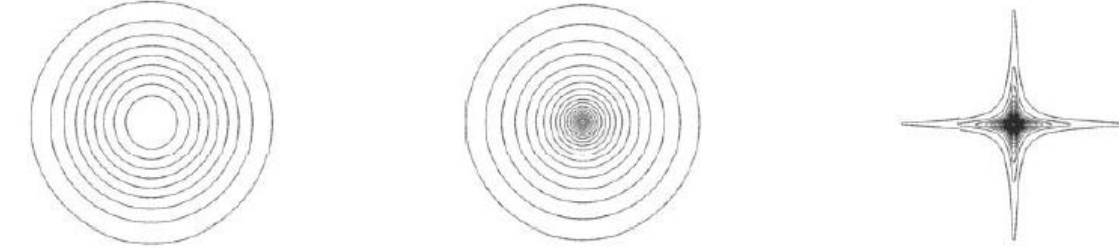
$K < M \ll N$

## Method

$$p(w_i | \alpha_i) = \prod_{j=1}^N \mathcal{N}(w_i(j) | 0, \alpha_{ij}^{-1})$$

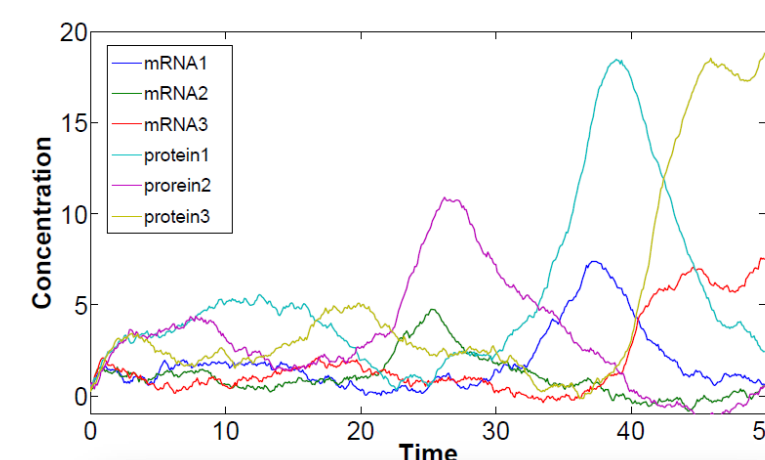


Gaussian prior      Marginal prior: single  $\alpha$       Independent  $\alpha$



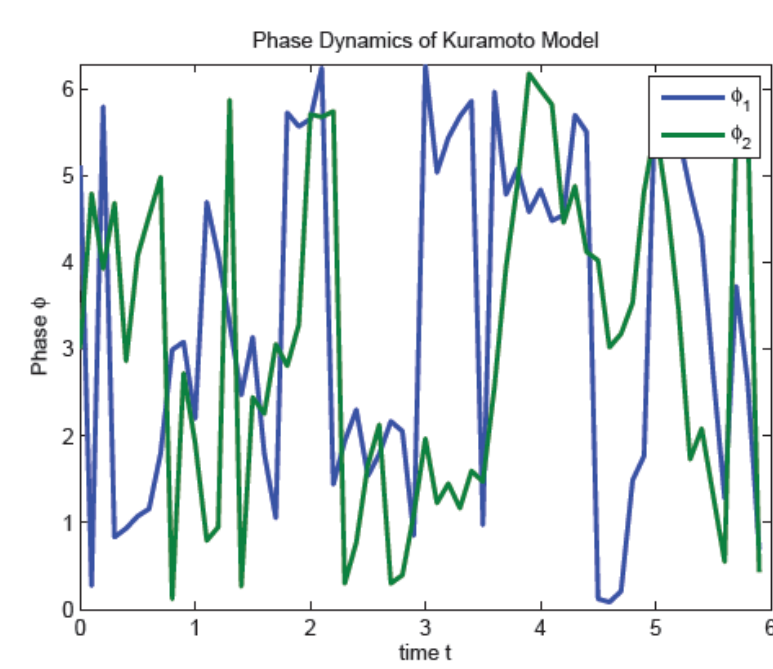
## Results

### Synthetic Genetic Repressilator



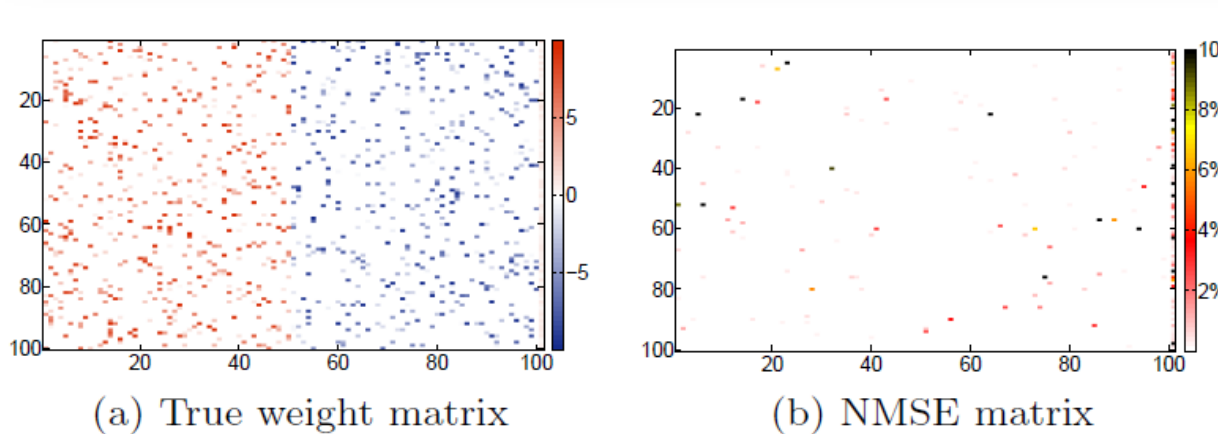
$$\begin{aligned} \frac{dx_1}{dt} &= -\gamma_1 x_1 + \frac{\alpha_1}{1 + x_6^{n_6}} + \theta_1, \\ \frac{dx_2}{dt} &= -\gamma_2 x_2 + \frac{\alpha_2}{1 + x_4^{n_2}} + \theta_2, \\ \frac{dx_3}{dt} &= -\gamma_3 x_3 + \frac{\alpha_3}{1 + x_5^{n_3}} + \theta_3, \\ \frac{dx_4}{dt} &= -\gamma_4 x_4 + \beta_1 x_1, \\ \frac{dx_5}{dt} &= -\gamma_5 x_5 + \beta_2 x_2, \\ \frac{dx_6}{dt} &= -\gamma_6 x_6 + \beta_3 x_3. \end{aligned}$$

### Kuramoto Oscillator



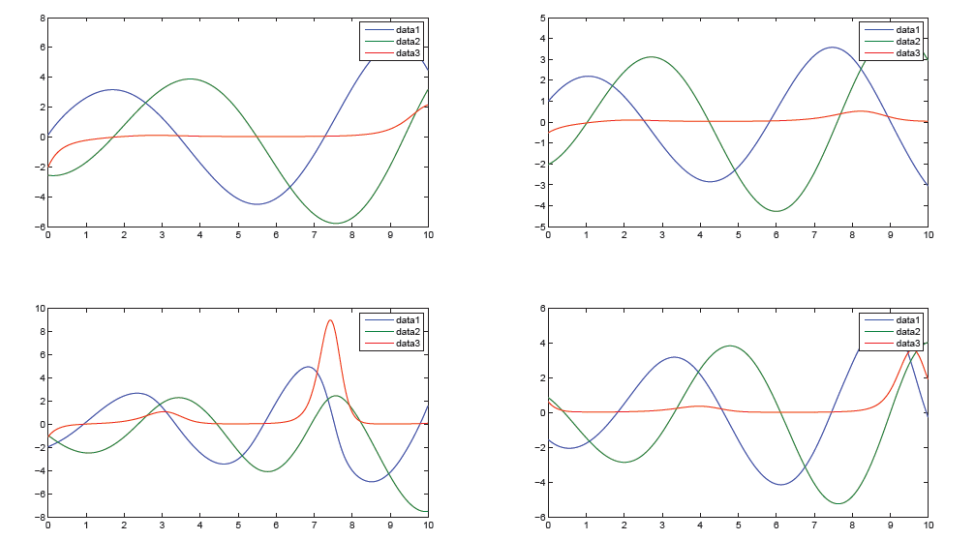
$$\dot{\phi}_i(t) = \omega_i + \sum_{j=1}^n w_{ij} g_{ij}(\phi_j(t) - \phi_i(t)) + \eta_i(t)$$

Phase      Natural Frequency      Coupling Function      Noise



## More results

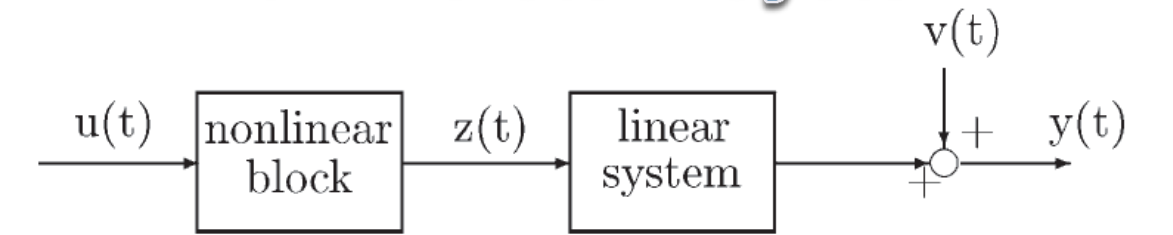
### Rössler Oscillator



$$\begin{aligned} \dot{p}_i(t) &= -\omega_i q_i(t) - s_i(t) + \sum_{j=1}^n w_{ij} g_{ij}^p(p_j(t) - p_i(t)) + \eta_i^p(t) \\ \dot{q}_i(t) &= \omega_i p_i(t) + a_i q_i(t) + \sum_{j=1}^n w_{ij} g_{ij}^q(q_j(t) - q_i(t)) + \eta_i^q(t) \\ \dot{s}_i(t) &= b_i + (p_i(t) - c_i) s_i(t) + \sum_{j=1}^n w_{ij} g_{ij}^s(s_j(t) - s_i(t)) + \eta_i^s(t) \end{aligned}$$

Local States      Local Function      Coupling Function      Noise

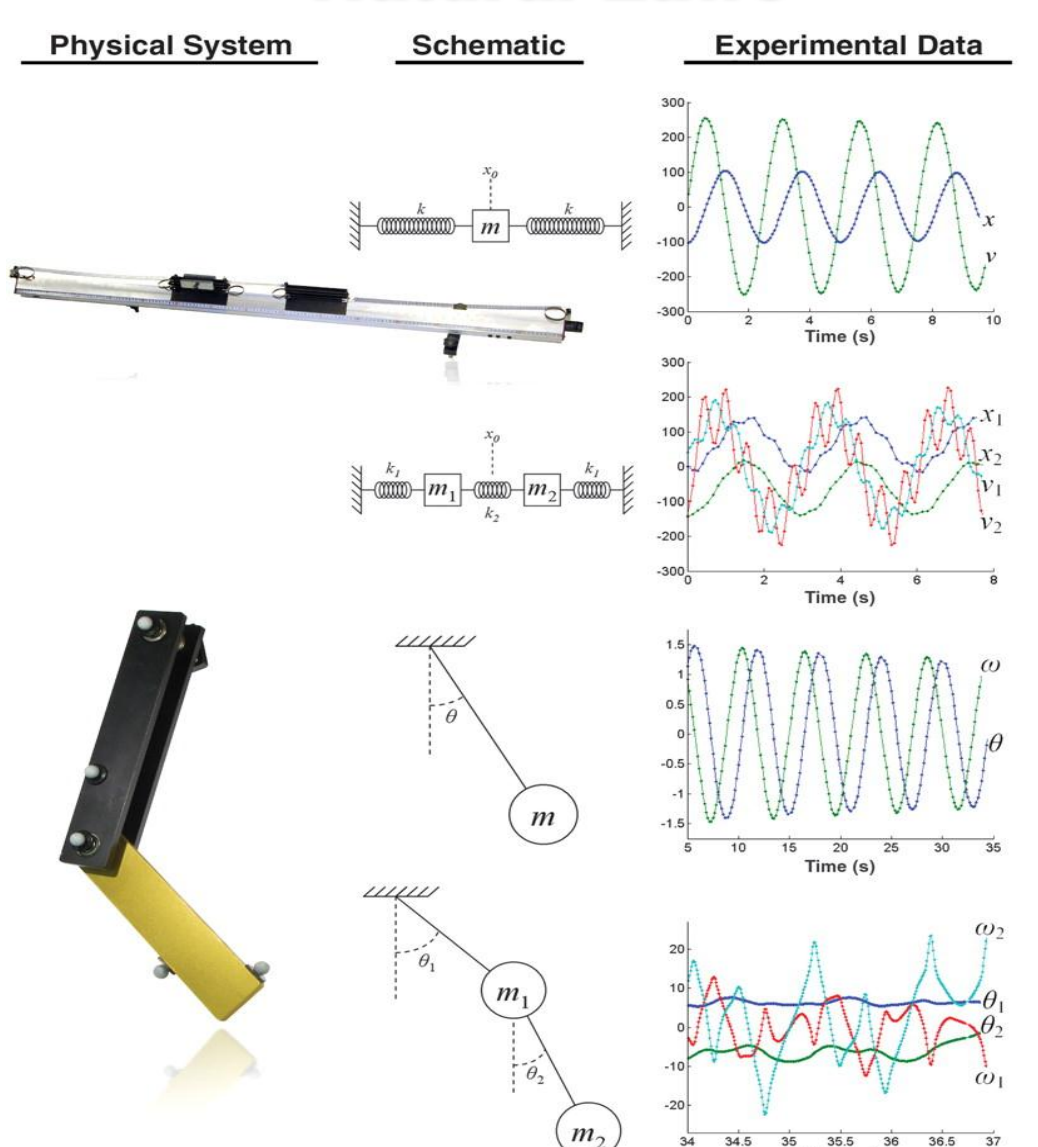
### Hammerstein System



$$y(t) = \sum_{j=1}^n b_j z(t-j) + v(t)$$

$$z(t) = \sum_{j=1}^m c_j g_j[u(t)]$$

### Natural Laws



Data from "Distilling Free-Form Natural Laws from Experimental Data", *Science* 324, 81 (2009)

## Conclusions

- The method requires time series data (noisy) and a series of candidate linear or nonlinear functions.
- Select a minimum number of candidate functions to fit the time series data.
- The network topology, functions with the corresponding parameters can be reconstructed.

## References

1. Pan W et al. *Control and Decision conference 2012*, submitted.
2. Pan W et al. *Physical Review Letters*, submitted.

## Acknowledgements

Acknowledge help from Dr Wei Dai, Imperial College London.

## Contact details

Wei Pan, w.pan11@imperial.ac.uk.

Presented at the Microsoft Research Summer School, July 2012