

# Inference for dynamics of continuous variables: the Extended Plefka Expansion with hidden nodes

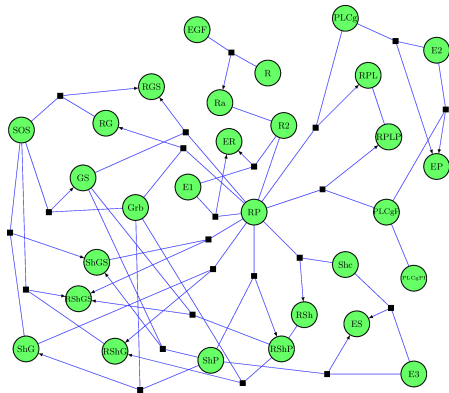
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NIPS, *Modelling and inference for dynamics workshop*,  
Montreal 11 December 2015

# Introduction and Motivation

Fundamental and practical **limitations**:



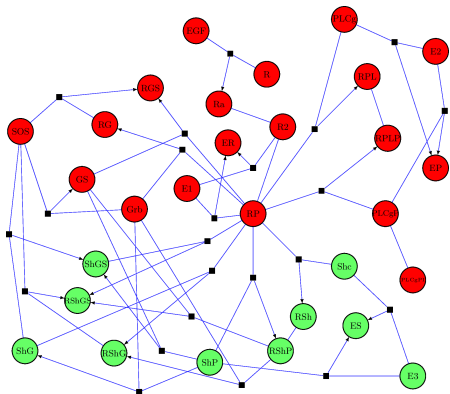
EGFR network from *Kholodenko et al. (1999)*

- Non-linear equations
- Vast amount of information ( $\sim 10^4$  eqs)
- **Uncertainty**  $\rightarrow$  No complete qualitative understanding

**SIZE, COMPLEXITY,  
UNCERTAINTY**



Statistical Physics for  
Model Reduction



Let us assume it is possible to characterize **only some** nodes

Small subset of variables:  
**Subnetwork**  $\rightarrow$  “observed”  
 Embedded in a larger network:  
**Bulk**  $\rightarrow$  “hidden” (unknown)

## Question

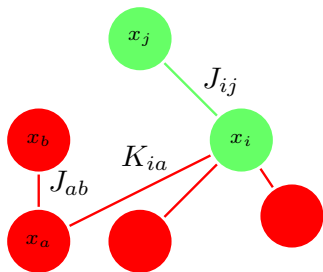
What can we say in general about the **INFERENCE** of hidden dynamics?

## LINEAR DYNAMICS

$i, j \rightarrow$  hidden (Bulk)

$a, b \rightarrow$  observed (Subnetwork)

$\xi_i, \xi_a$  Gaussian white noises



$$\langle \xi_i(t) \xi_j(t') \rangle = \sigma_b^2 \delta_{ij} \delta(t - t')$$

$$\langle \xi_a(t) \xi_b(t') \rangle = \sigma_s^2 \delta_{ab} \delta(t - t')$$

$$\partial_t x_i(t) = -\lambda x_i(t) + \sum_j J_{ij} x_j(t) + \sum_a K_{ia} x_a(t) + \xi_i(t)$$

$$\partial_t x_a(t) = -\lambda x_a(t) + \sum_b J_{ab} x_b(t) + \sum_j K_{aj} x_j(t) + \xi_a(t)$$

# Extended Plefka Expansion

B. Bravi, P. Sollich, M. Opper, J. Phys. A., submitted, arXiv:1509.07066 (2015)

## DYNAMICAL MEAN FIELD APPROXIMATION

- 1 Effectively **non-interacting** dynamics:  
Couplings are replaced by **memory** + **coloured** noise
- 2 Extension: 1<sup>st</sup> and 2<sup>nd</sup> moments are constrained
- 3 **Gaussian** statistics conditioned on observations

**Posterior mean**  $\mu_i(t)$  = best estimate of hidden dynamics  
**Forward-Backward** propagation

**Posterior variance**  $C_i(t, t)$  = estimate of **error** of prediction

Expect Plefka to give exact  $\mu_i(t)$  and  $C_i(t, t)$  (=errors) when **Mean Field** interactions + **Thermodynamic limit**  $N^B \rightarrow \infty$

**Stationary Regime**  $\rightarrow$  Time Translation Invariant

Average  $\tilde{C}^B(\omega) = \frac{1}{N^B} \sum_i \tilde{C}_i(\omega)$  in **Fourier** space

$$\tilde{C}^B(\omega) = \underbrace{\frac{\sigma_s^2}{k^2}}_{\text{Amplitude}} \underbrace{C_{\alpha, \gamma, \eta, p}^B(\Omega)}_{\text{Dimensionless}}$$

## Dimensionless parameters

- $\alpha = \frac{N^S}{N^B}$  Ratio observations/hidden states
- $\gamma = \frac{j}{\lambda}$  Stability of hidden dynamics  $\gamma < \gamma_c = \frac{1}{1+\eta}$
- $p = \frac{\lambda}{\sigma}$  Decay constant & hidden-to-observed coupling

$\sigma = \frac{\sigma_b k}{\sigma_s}$  Defines the frequency scale:  $\Omega = \frac{\omega}{\sigma}$

# Critical regime

We analyze in the parameter space  $\alpha, \gamma, p$  the **singularities** of  $\mathcal{C}^B(0)$

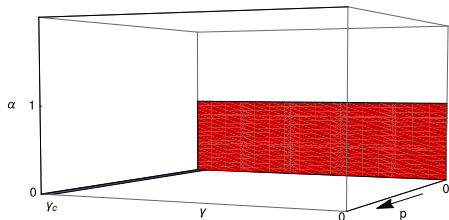
## Critical regions

①  $\forall p, \alpha = 0$  and  $\gamma > \gamma_c$

No observations: internal stability

②  $\forall \gamma, p = 0$  and  $0 < \alpha < 1$   
(i.e.  $k \gg \lambda$  at fixed  $\frac{\sigma_s}{\sigma_b}$ )

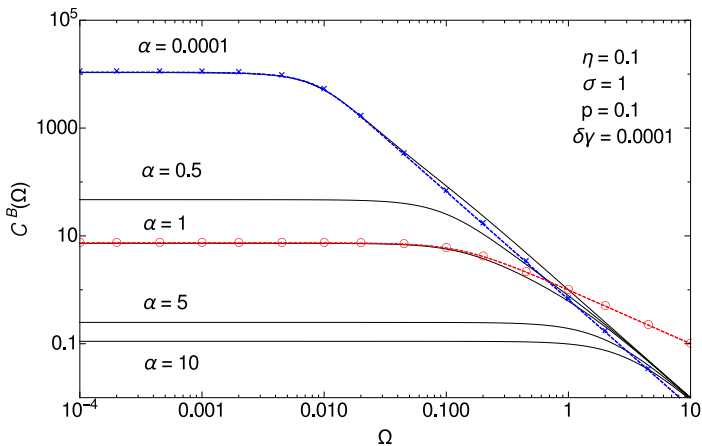
**“Underconstrained”** hidden system: strong constraints from observations, but too few



Power-law dependence of  $C^B(0)$  on  $\delta\alpha$ ,  $\delta\gamma$ ,  $p$ :

**Scaling analysis**  $\Rightarrow$  **Master curves**

Information on **relaxation** times and inference **error**



$$\Omega^* \ll \Omega \ll 1 \quad \alpha \rightarrow 0 \quad C^B \sim \frac{1}{\Omega^2} \quad \alpha \rightarrow 1 \quad C^B \sim \frac{1}{\Omega}$$



Thank you for your attention!