

# Efficient inference of interactions from non-equilibrium data and application to multi-electrode neural recordings

ESR: Claudia Battistin<sup>1</sup>  
Supervisor: Dr. Yasser Roudi<sup>1,2</sup>

<sup>1</sup>Kavli Institute for Systems Neuroscience, Trondheim (NO)

<sup>2</sup>NORDITA, Stockholm (SE)

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NETADIS MID-TERM REVIEW  
Institut Henri Poincaré, Paris



Kavli Institute for Systems Neuroscience  
Centre for Neural Computation



- 1 My background
- 2 Training experiences
- 3 My work so far
- 4 Future developments and expectations

July 2012 Master degree in theoretical physics (University of Trieste)

Master's Thesis: *"Criticality of models inferred in Boltzmann learning"*

Supervisor: Dr. Matteo Marsili



- **September 2012** start as NETADIS ESR at the Kavli Institute for Systems Neuroscience, Trondheim NORWAY
- **October 2012** enrolled as PhD Student in Neuroscience at NTNU, Trondheim NORWAY



# Training experiences

## 2012

October	Random Matrix Theory by PierPaolo Vivo	Kavli I., Trondheim
11/26-12/07	Winter School on Quantitative Systems Biology	ICTP, Trieste

## 2013

02/03-02/06	Netadis First Scientific Meeting	Torino
02/13-02/22	Spin glasses by Adriano Barra	NORDITA, Stockholm
03/07-03/15	Stochastic Thermodynamics (Workshop)	NORDITA, Stockholm
04/03-04/05	SMED8005 Communication of Science	Kavli I., Trondheim
05/20-06/14	Spring College on Physics of Complex Systems	ICTP, Trieste
08/09-08/19	Large Deviations by prof.Kamiar Rahnama Rad	Kavli I., Trondheim
09/08-09/22	Netadis Summer School 2013	Hillerod, Denmark
23/10-20/12	Secondment at TUB in prof.Opper group	Berlin

- 1 Path integral Methods:  
One-loop correction to the Laplace Approximation for systems out of equilibrium

*in collaboration with prof. K. Rahnama Rad (Columbia University, NY)*

- 2 Gaussian Average Method for the kinetic Ising Model

- 3 Variational Factorizable Approximation for the kinetic Ising Model

*under the supervision of prof. M.Opper & Dr. Y.Roudi*

## 1 Path integral Methods:

One-loop correction to the Laplace Approximation for systems out of equilibrium

- TAP:  $m_i(t) = \tanh \left[ \sum_j J_{ij} m_j(t-1) - m_i(t) \sum_j J_{ij}^2 (1 - m_j(t-1)^2) \right]$  (Pleška expansion)

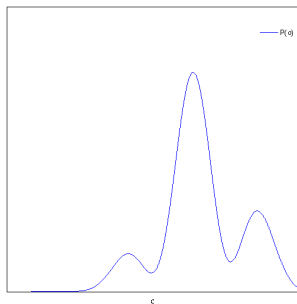
*in collaboration with prof. K. Rahnama Rad (Columbia University, NY)*

## 2 Gaussian Average Method for the kinetic Ising Model

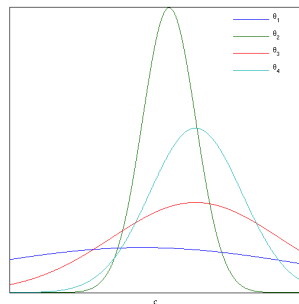
## 3 Variational Factorizable Approximation for the kinetic Ising Model

*under the supervision of prof. M. Opper & Dr. Y. Roudi*

## Variational approximation



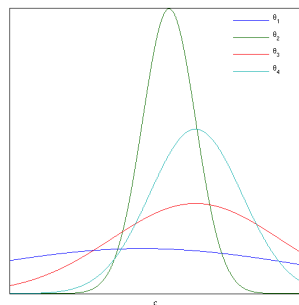
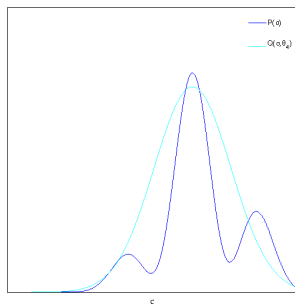
model distribution  $P(\sigma)$



ensemble  $Q(\sigma, \theta)$



## Variational approximation



optimal approximate distribution  
 $Q(\sigma, \theta^*)$  (TREATABLE!)

ensemble  $Q(\sigma, \theta)$



KL-divergence minimization

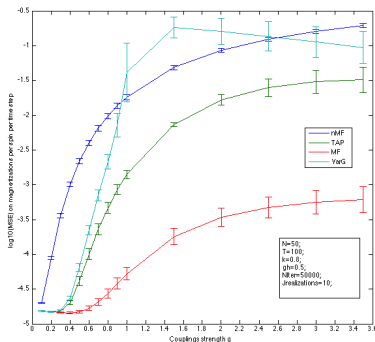
## Gaussian Average Method for the kinetic Ising Model

$$\bullet m_i(t) = \int \frac{D\mathcal{X}}{(2\pi)^{NT}} \tanh \left[ x_i(t-1) + \sum_j J_{ij} m_j(t-1) \right]$$

$$J_{ij} \perp J_{ji} \rightarrow \text{Mezard's MF: } m_i(t) = \int D\mathbf{x} \tanh \left[ \sum_j J_{ij} m_j(t-1) + x \sqrt{\Delta_i(t-1)} \right]$$

where  $\Delta_i(t) = \sum_j J_{ij}^2 (1 - m_j(t)^2)$

$$g \ll 1 \rightarrow \text{TAP: } m_i(t) = \tanh \left[ \sum_j J_{ij} m_j(t-1) - m_i(t) \sum_j J_{ij}^2 (1 - m_j(t-1)^2) \right]$$



$k = 0.8$

## Variational Factorizable Approximation for the kinetic Ising Model

### Equilibrium Ising model

- Saddle-point Approximation  $\searrow$
  - Variational Factorizable Approximation  $\nearrow$
- Naïve Mean Field Equations

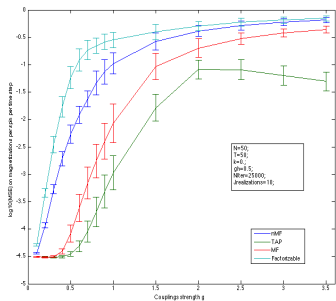
### Kinetic Ising model

- Saddle-point Approximation  $\rightarrow m_i(t) = \tanh \left[ \sum_j J_{ij} m_j(t-1) \right]$  (nMF)
- Variational Factorizable Approximation  $\rightarrow m_i(t) = \tanh \left[ \sum_k J_{ik} m_k(t-1) + \sum_k J_{ki} m_k(t+1) - \sum_{j \neq i} \left\langle \tanh^{-1} \left[ \tanh \left[ \sum_{k \neq i} J_{jk} \sigma_k(t) \right] \tanh [J_{ji}] \right] \right\rangle_Q \right]$   
where  $Q(\sigma)$  is the factorizable distribution

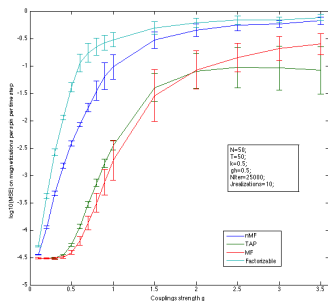
Why poorer predictions?

$$D_f(\mathbf{m}, \mathbf{H}) \geq D_{nMF}(\mathbf{m}, \mathbf{H})$$

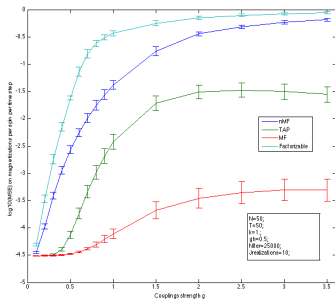
# My work so far



$k=0$



$k=0.5$



$k=1$



**February 2014**

Dilute systems:  
The Bethe Approximation for the kinetic Ising Model

*supervised by prof. John Hertz (NORDITA, Copenhagen) & Dr. Y. Roudi*

**March 2014**

Deep learning in networks with hidden nodes

*in collaboration with Benjamin Dunn*

**Summer 2014**

Learning the connectivity of the brain:  
performances of the approximations developed so far

*using data available at the Kavli Institute for Systems Neuroscience*

**dates to be decided**

Secondment in London at prof. Sollich group

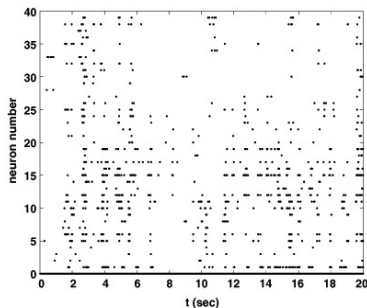
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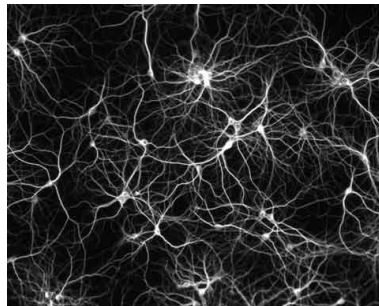
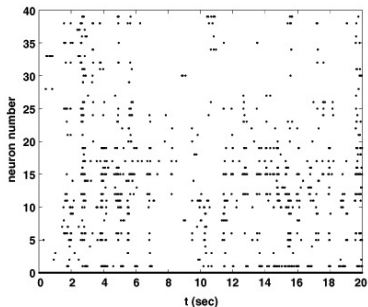
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Thanks!