# Mean-field method with correlations determined by linear response

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#### Outline

- Mean field approx. (MFA)
- Linear response (LR)
- Limits of MFA and LR
- General framework for MFA + LR
- Numerical results on 2D models (both ordered and with disorder)

#### The main pb. in Stat Mech

• Compute the free-energy

 $F(\boldsymbol{J},\boldsymbol{h}) = \log Z(\boldsymbol{J},\boldsymbol{h}) = \log \sum_{\{s_i\}} \exp\left(\sum J_{ij} s_i s_j + \sum_i h_i s_i\right)$ 

The sum is over exponentially many terms

- Resort to mean field approximations (MFA)
  - few parameters to be fixed self-consistently
  - fast to compute, but inexact!

naive mean-field (nMF) only N parameters  $m_i = \langle s_i \rangle$  $H(x) \equiv -x \ln(x)$  $F_{\rm nMF} = \sum_{i} \left[ H\left(\frac{1+m_i}{2}\right) + H\left(\frac{1-m_i}{2}\right) \right] + \sum_{i} h_i m_i + \sum_{i \neq i} J_{ij} m_i m_j$  $\frac{\partial F_{\rm nMF}}{\partial m_i} = \sum_j J_{ij} m_j + h_i - \operatorname{atanh}(m_i) = 0$  $m_i = \tanh \left| h_i + \sum_j J_{ij} m_j \right|$ 

• nMF + Onsager reaction term (TAP)

$$F_{\text{TAP}} = \sum_{i} \left[ H\left(\frac{1+m_{i}}{2}\right) + H\left(\frac{1-m_{i}}{2}\right) \right] + \sum_{i} h_{i}m_{i} + \sum_{i \neq j} \left( J_{ij}m_{i}m_{j} + \frac{1}{2}J_{ij}^{2}(1-m_{i}^{2})(1-m_{j}^{2}) \right)$$

$$m_{i} = \tanh \left[ h_{i} + \sum_{j} J_{ij} \left( m_{j} - J_{ij} (1 - m_{j}^{2}) m_{i} \right) \right]$$
reaction term

Expansion in small J (correct at high temperature or in fully connected models)

$$F_{\rm nMF} = \sum_{i} \left[ H\left(\frac{1+m_i}{2}\right) + H\left(\frac{1-m_i}{2}\right) \right] + \sum_{i} h_i m_i + \sum_{i \neq j} J_{ij} m_i m_j$$

$$F_{\text{TAP}} = \sum_{i} \left[ H\left(\frac{1+m_{i}}{2}\right) + H\left(\frac{1-m_{i}}{2}\right) \right] + \sum_{i} h_{i}m_{i} + \sum_{i \neq j} \left( J_{ij}m_{i}m_{j} + \frac{1}{2}J_{ij}^{2}(1-m_{i}^{2})(1-m_{j}^{2}) \right) \right]$$

• Bethe approximation (BA) nMF + nearest neighbors correlations parameters  $c_{ij} = \langle s_i s_j \rangle - m_i m_j$ 

$$\begin{split} F_{\mathrm{BA}} &= \sum_{i \neq j} \left[ H\left(\frac{(1+m_i)(1+m_j)+c_{ij}}{4}\right) + H\left(\frac{(1-m_i)(1-m_j)+c_{ij}}{4}\right) + \right. \\ &+ H\left(\frac{(1+m_i)(1-m_j)-c_{ij}}{4}\right) + H\left(\frac{(1-m_i)(1+m_j)-c_{ij}}{4}\right) \right] + \\ &+ \sum_i (1-d_i) \left[ H\left(\frac{1+m_i}{2}\right) + H\left(\frac{1-m_i}{2}\right) \right] + \sum_i h_i m_i + \sum_{i \neq j} J_{ij}(c_{ij}+m_i m_j) \;, \end{split}$$

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• Bethe approximation (BA)

$$f(m_1, m_2, t) = \frac{1 - t^2 - \sqrt{(1 - t^2)^2 - 4t(m_1 - m_2 t)(m_2 - m_1 t)}}{2t(m_2 - m_1 t)}$$

$$m_i = \tanh\left[h_i + \sum_j \operatorname{atanh}\left(t_{ij}f(m_j, m_i, t_{ij})\right)\right]$$

$$c_{ij}(m_i, m_j, t_{ij}) = \frac{1}{2t_{ij}} \left( 1 + t_{ij}^2 - \sqrt{(1 - t_{ij}^2)^2 - 4t_{ij}(m_i - t_{ij}m_j)(m_j - t_{ij}m_i)} \right) - m_i m_j$$
$$t_{ij} = \tanh(J_{ij})$$

## Computing correlations by linear response (LR)

- Correlations are trivial in MFA  $C_{ij} = 0$  in nMF, TAP and BA (between distant spins)
- Non trivial correlations can be obtained by using the linear response

$$\chi_{ij} = \frac{\partial m_i}{\partial h_j} \qquad (\chi^{-1})_{ij} = \frac{\partial h_i}{\partial m_j}$$

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$$(\chi_{\rm nMF}^{-1})_{ij} = \frac{\delta_{ij}}{1 - m_i^2} - J_{ij} ,$$
  
$$(\chi_{\rm TAP}^{-1})_{ij} = \left[\frac{1}{1 - m_i^2} + \sum_k J_{ik}^2 (1 - m_k^2)\right] \delta_{ij} - (J_{ij} + 2J_{ij}^2 m_i m_j)$$

## Computing correlations by linear response in BA

 Analytic expression for the correlations in BA (FRT, JSTAT, 2012)

$$(\chi_{\text{BA}}^{-1})_{ij} = \left[\frac{1}{1-m_i^2} - \sum_k \frac{t_{ik}f_2(m_k, m_i, t_{ik})}{1-t_{ik}^2 f(m_k, m_i, t_{ik})^2}\right] \delta_{ij} - \frac{t_{ij}f_1(m_j, m_i, t_{ij})}{1-t_{ij}^2 f(m_j, m_i, t_{ij})^2}$$

- Coincide with the fixed point of Susceptibility Propagation (no need to run any algorithm!)
- PROBLEM: estimates of nearest neighbor (NN) correlations are inconsistent  $\chi^{\rm BA}_{ij} 
  eq C^*_{ij}$

### How loops make MFA fail

e.g. Bethe approximation, high temperature phase

• maximum entropy (free energy minimum)

$$\langle \sigma_i \sigma_j \rangle_c^{\mathrm{BA}} = c_{ij}^* = \tanh(\beta J_{ij}) < \langle \sigma_i \sigma_j \rangle_c^{\mathrm{exact}}$$

linear response

$$\chi_{ii} = 1 + \sum_{j \in \partial i} u_{j \to i, i} \neq 1$$



ferromagnet  $\chi_{ij}^{\rm BA} > \langle \sigma_i \sigma_j \rangle_c^{\rm exact}$ 

#### Make MFA & LR consistent

"Consistency is more important than truth" (S. Ting)

Add to the free-energy Lagrange multipliers to enforce (at the fixed point)



#### General framework (MFA + LR)



#### General framework (MFA + LR)

$$energy$$

$$F_{MFA}(\{m_i\}, \{C_{ij}\}, ...) = \underbrace{\sum_{i} h_i m_i + \sum_{ij} J_{ij}(m_i m_j + C_{ij})}_{ij} + \underbrace{\sum_{i} \left[ H\left(\frac{1+m_i}{2}\right) + H\left(\frac{1-m_i}{2}\right) \right] + S^{2+}(\{m_i\}, \{C_{ij}\}, ...)}_{entropy}$$

#### General framework (MFA + LR)

$$\begin{split} F_{\lambda} &= F_{\mathrm{MFA}}(\{m_i\}, \{C_{ij}\}, \ldots) + \sum_{i} \lambda_i m_i^2 + \sum_{i < j} \lambda_{ij} C_{ij} \\ & \text{Maximum Entropy equations} \\ & \text{from free-energy minimization} \\ \partial_{m_i} F_{\lambda} &= \operatorname{atanh}(m_i) + \partial_{m_i} S^{2+} + h_i + \sum_{j} J_{ij} m_j + \lambda_i m_i = 0 \\ \partial_{C_{ij}} F_{\lambda} &= \partial_{C_{ij}} S^{2+} + J_{ij} + \lambda_{ij} = 0 \\ \partial_{C_{ijk}} F_{\lambda} &= \partial_{C_{ijk}} S^{2+} = 0 \\ \end{split}$$

 $\chi_{ii} = 1 - m_i^2 \quad \chi_{ij} = C_{ij}$ 

#### Some comments...

- Bethe/CVM free-energies are not convex adding parameters it not obvious to improve!
- $\lambda$  parameters measure how wrong is the MFA e.g. on a random graph with Bethe approx.  $\lambda \to 0$
- $J_{ij} \rightarrow J_{ij} + \lambda_{ij}$  would naively imply stronger correlations and also more unstable Susc. Prop. but this is not the case!

#### Models studied so far

- Ising variables
- 2D topology (square and triangular lattices)
- ferromagnetic
- frustrated -> triangular antiferromagnet
- frustrated & disordered -> random field

#### Nearest-neighbor correlation (2D square lattice)



#### Nearest-neighbor correlation (2D triangular lattice)



2D RFIM <h> = 0.0  $\sigma_h$  = 0.2  $\beta$  = 0.25



2D RFIM  $<h> = 0.0 \sigma_h = 0.2 \beta = 0.25$ 



2D RFIM  $\langle h \rangle = 0.0 \ \sigma_h = 0.2 \ \beta = 0.25$ 



2D RFIM  $<h> = 0.0 \sigma_h = 0.2 \beta = 0.25$ 



#### Random field Ising model 2D square lattice $\langle h \rangle = 0.2 \ \sigma_h = 0.5$



### Summary

Any MFA (even Cluster Variational Methods) can be improved by enforcing consistency between maximum entropy and linear response estimates.

Several improvements achieved

- Direct problem: better high temperature expansions
   & better correlation estimates
- Inverse problem: smaller errors in estimating couplings and fields

Still working for a fast message passing algorithm...