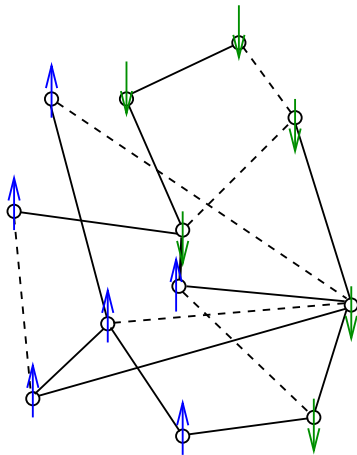


Spin glasses on locally tree like graphs

Amir Dembo, Antoine Gerschenfeld, Andrea Montanari

BIRS, Banff, July 22, 2014

Ising spin glass



$G_n = (V_n \equiv [n], E_n)$ finite, undirected graphs.

$x_i \in \{+1, -1\}$

$$\mu(\underline{x}) = \frac{1}{Z_{n,J}(\beta, B)} \exp \left\{ \beta \sum_{(ij) \in E_n} J_{ij} x_i x_j + B \sum_{i=1}^n x_i \right\}$$

$J_{ij} \in \{+1, -1\}$ uniformly random

[Viana, Bray 1985]

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Locally tree-like graphs (GW case)

$P \equiv \{P_k\}_{k \geq 0}$	Degree distribution, law of L (of mean $\bar{P} > 0$)
$\rho \equiv \{\rho_k\}_{k \geq 0}$	Size-biased P , law of K (degree of uniform edge)
$T(P, \rho, t)$	t -generations MGW tree (root degree P , else ρ)
$B_i(t)$	Ball of radius t in G_n centered at node i

Definition

$\{G_n\}$ converges locally to $T(P, \rho, \infty)$ if for uniformly random $I \in [n]$ and fixed t , law of $B_I(t)$ converges as $n \rightarrow \infty$ to $T(P, \rho, t)$.
 $\{G_n\}$ uniformly sparse if $\{|\partial I|\}$ is uniformly integrable.

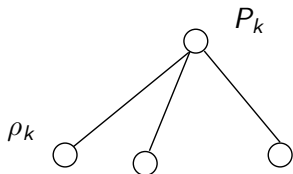
[Benjamini & Schramm 2001, c.f. Aldous & Lyons 2007]

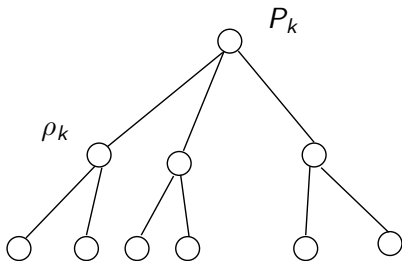
MGW $T(P, \rho, t)$

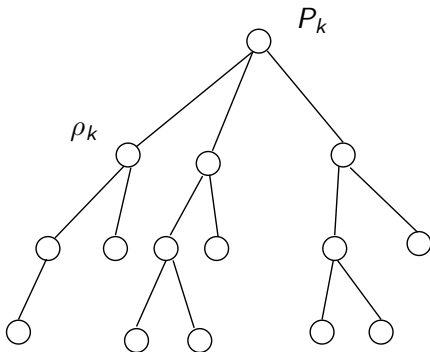
P_k

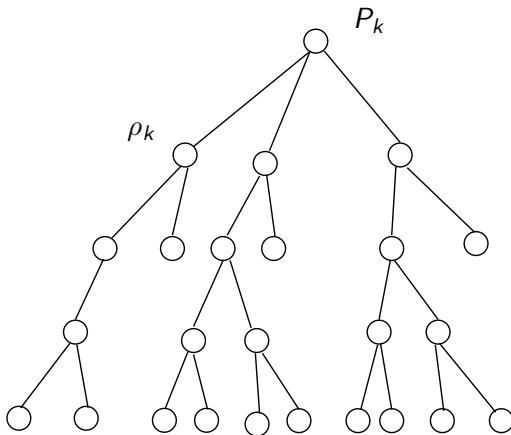


ρ_k









$\mu_{i \rightarrow j}(\cdot)$ \equiv Marginal of x_i when replacing $e^{\beta J_{ij} x_i x_j}$ by 1

$$\mu_{i \rightarrow j}(x_i) \approx \frac{1}{Z_{i \rightarrow j}} e^{B x_i} \prod_{l \in \partial i \setminus j} \sum_{x_l} e^{\beta J_{il} x_i x_l} \mu_{l \rightarrow i}(x_l)$$

Philosophy: approximate **local marginals** of $\mu(\cdot)$ in terms of measures on trees.

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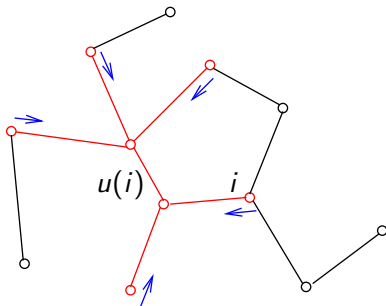
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Bethe-Peierls approximation



$F = (U, E_U) \subseteq G$, $\text{diam}(F) \leq t$, such that $\partial i \in U$ or $\partial i \cap U = \{u(i)\}$

$$\mu_U(\underline{x}_U) \approx \nu_U(\underline{x}_U) = \frac{1}{Z_U} \prod_{i \in U \setminus \partial U} e^{Bx_i} \prod_{(i,j) \in E_U} e^{\beta J_{ij} x_i x_j} \prod_{i \in \partial U} \nu_{i \rightarrow u(i)}(x_i).$$

$\{\nu_{i \rightarrow j}(\cdot)\}$ \rightarrow 'set of messages' (aka cavity fields)

Bethe free energy density

For iid $\theta_i \in \{+\tanh \beta, -\tanh \beta\}$ uniformly at random, independent of K, L and iid $m_i^{(t)}$, let

$$m^{(t+1)} \stackrel{d}{=} \tanh \left\{ B + \sum_{i=1}^{K-1} \operatorname{atanh}(\theta_i m_i^{(t)}) \right\}, \quad m^{(0)} = 0.$$

$$\beta < \beta_c(B, P) \quad \Rightarrow \quad m^{(t)} \xrightarrow{d} m^*$$

m^* fixed point of **Bethe** or **belief propagation** recursion (on message sets)

$$\beta_c(0, P) = \operatorname{atanh}(1/\sqrt{P}).$$

For iid m_i^* independent of L , $\gamma = \tanh(B)$,

$$\phi_*(P, \beta, B) = \phi^V - \phi^E$$

$$\phi^V = \log \cosh B + \mathbb{E} \log \left\{ (1 + \gamma) \prod_{i=1}^L (1 + \theta_i m_i^*) + (1 - \gamma) \prod_{i=1}^L (1 - \theta_i m_i^*) \right\}$$

$$\phi^E = \frac{\bar{P}}{2} \mathbb{E} \log(1 + \theta_0 m_1^* m_2^*) - \frac{\bar{P}}{2} \log \cosh \beta.$$

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Free energy density: brief survey

$$\phi = \lim_{n \rightarrow \infty} \phi_{n, \mathbf{J}} \quad \phi_{n, \mathbf{J}} = \frac{1}{n} \log Z_{n, \mathbf{J}}(\beta, B).$$

Theorem

If G_n uniformly sparse, converges locally to $\mathbb{T}(P, \rho, \infty)$ and $\beta < \beta_*(B, P)$, then $\phi = \phi_*(P, \beta, B)$.

Uniformly random G_n , average degree γ

$B = 0$, $\beta_* = \operatorname{atanh}(1/\sqrt{\gamma})$ Guerra, Toninelli 2003

$B \neq 0$, $\beta_* = O(1/\gamma)$ Talagrand 2001, 2003.

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$\beta_*(0, P) = \operatorname{atanh}(1/\sqrt{\bar{\rho}})$; $K \sim k_{\text{typ}} \gg 1 \Rightarrow \beta_*(B, P) \simeq \frac{f(B)}{\sqrt{k_{\text{typ}}}}$ and

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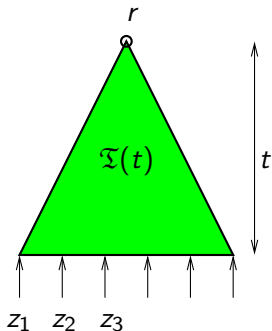
1. Reduce to expectations of local quantities

$$\frac{d\phi_{n,\mathbf{J}}}{d\beta} = n^{-1} \sum_{(ij) \in E_n} \langle x_i x_j \rangle_{n,\mathbf{J}}$$

($\langle \cdot \rangle_{n,\mathbf{J}}$ is expectation under Ising spin glass \mathbf{J} on G_n).

2. Prove convergence of local expectations to tree values.

Convergence to tree values



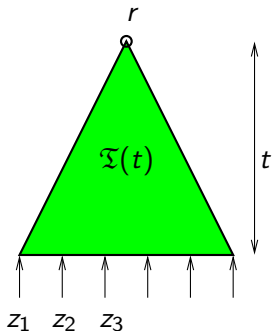
\mathfrak{T} infinite tree with max degree k_{\max}

$\mathfrak{T}(t)$ first t generations

$\mu_{j,z}^{t,z}(\cdot)$ Ising spin glass on $\mathfrak{T}(t)$ boundary condition z

$\mu_{j,r}^{t,z}$ root spin expectation

Convergence to tree values



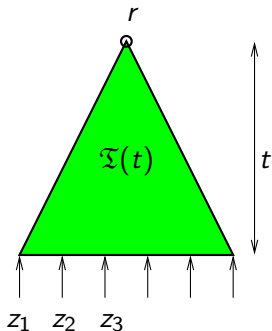
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Decorrelation on random tree at high temperature

Uniform (Gibbs measure uniqueness)

$$\mathbb{E}_{\mathbf{T}} \mathbb{E}_{\mathbf{J}} \sup_{z(1), z(2)} |\mu_{\mathbf{J}, r}^{t, z(1)} - \mu_{\mathbf{J}, r}^{t, z(2)}| \rightarrow 0$$

$$\beta = O(1/k_{\text{typ}})$$

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$$\mathbb{E}_T \sup_z \mathbb{E}_J |\mu_{J,r}^{t,z} - \mu_{J,r}^{t,\text{free}}| \leq \eta(\beta, B)^t \rightarrow 0$$

$$\eta(\beta, B) < 1 \text{ at } \beta \simeq \frac{f(B)}{\sqrt{k_{\text{typ}}}}$$

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