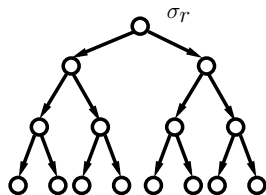


Decay of Correlations for the Hardcore Model in Random Regular Graphs

Nayantara Bhatnagar (University of Delaware)

joint work with

Allan Sly (UC Berkeley) and Prasad Tetali (Georgia Tech)

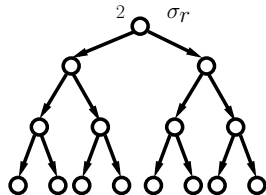


Broadcast Model

- Infinite d -ary tree T_d .
- Spins $[q]$, $\sigma_v \in [q]$ is spin at v .
- Choose $\sigma_r \sim \pi$.
- Independent Markov chain M on each edge of T .

Motivation

- Evolution of DNA, Noisy computation.
- Broadcast model gives rise to the **free Gibbs measure**.
- Random constraint satisfaction.

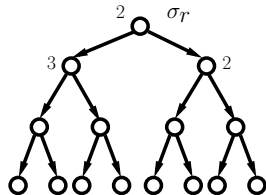


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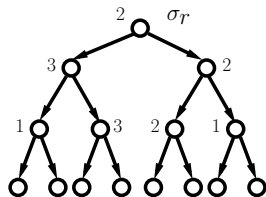
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Information Flow on Trees

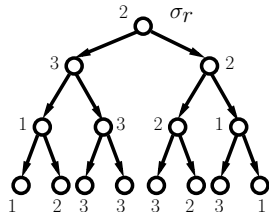


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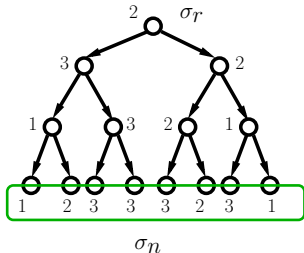
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What is reconstruction?

- Reconstruction is **solvable** if for some i, j

$$\limsup_n d_{tv}(\sigma_n^i, \sigma_n^j) > 0$$

- Mutual information between values at level n and root does not go to 0.

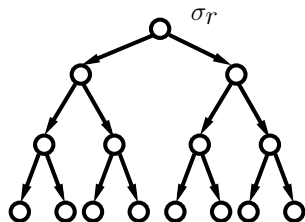
$$\mathbb{P}(\sigma_r = i | \sigma_n) \not\rightarrow \pi_i \quad \text{a.s.}$$

- Otherwise, **non-solvable** or **non-reconstruction**.

Related to...

- Can the phylogenetic tree be constructed efficiently from DNA?
- Extremality of the free Gibbs measure.
- Phase transitions in the solution space of random constraint satisfaction problems.

Symmetric channel



- Keep state w.p. λ , randomize w.p., $1 - \lambda$.
- Second eigenvalue $\lambda(M) = \lambda$.

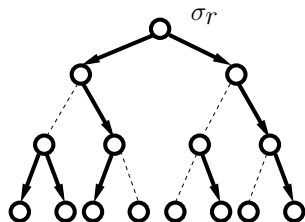
$$M_{i,j} = \lambda \mathbb{1}_{i=j} + \frac{1 - \lambda}{q}$$

- Critical λ_c s.t.
 - Reconstruction solvable for $\lambda > \lambda_c$.
 - Non-solvable for $0 < \lambda < \lambda_c$.

Percolation bound

Non-solvable when $\lambda d \leq 1$. All components finite a.s.

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-

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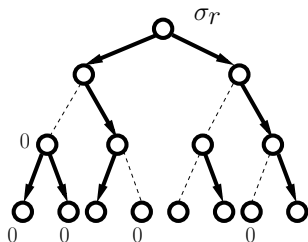
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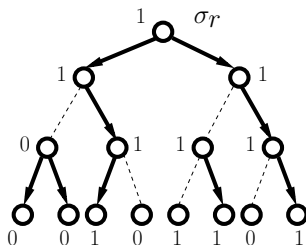
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Non-solvable when $\lambda d \leq 1$. All components finite a.s.

Theorem (Kesten-Stigum '68)

For $\lambda^2 d > 1$, reconstruction is (census) solvable.

Theorem (Mossel-Peres '03)

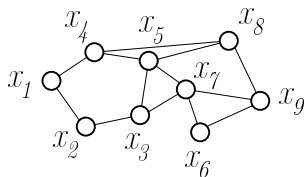
The threshold for census reconstruction is $\lambda^2 d = 1$.

- *Kesten-Stigum CLT's imply a.s. the frequencies are π .*

Tightness of Kesten-Stigum

- [Bleher-Ruiz-Zagrebnoy '95], [Ioffe '96], [Evans-Kenyon-Peres-Schulman '00] K-S bound tight for binary symmetric channel ($q = 2$, Ising model).
- [Mossel '01] When $q \geq 18$, K-S bound is **not** tight. Reconstruction beyond census.
- [Mezard-Montanari '06] Statistical physics predictions: for $\theta > 0$ (ferromagnetic)
 - K-S bound tight when $q \leq 4$.
 - K-S bound **not** tight for $q \geq 5$.
- [Sly '08] rigorously verified these predictions, for large enough d .
- [Brightwell-Winkler '04] Reconstruction beyond K-S in the "hardcore model" of lattice gases.

Graphical models



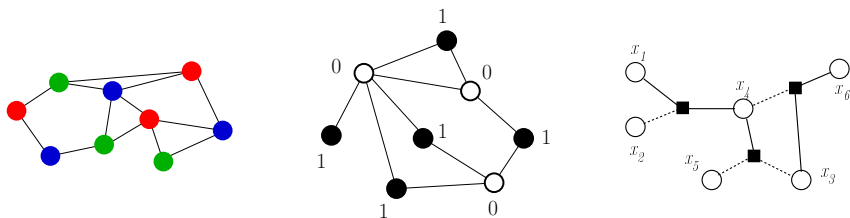
$G = (V, E)$, $|V| = n$, $\underline{x} = (x_1, \dots, x_n)$, $x_i \in [q]$, $\psi : [q] \times [q] \rightarrow \mathbb{R}$.

$$\mu(\underline{x}) = \frac{1}{Z} \prod_{(x_i \sim x_j)} \psi(x_i, x_j)$$

- G has bounded degree.
- G is random.
- G has girth $g(n) \rightarrow \infty$, except for $o(n)$ vertices.

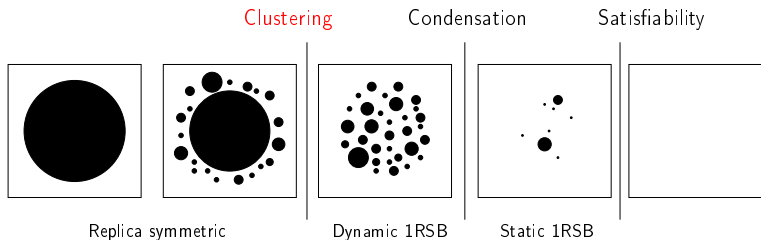
Random constraint satisfaction problems

A CSP is a set of variables subject to some constraints: decide whether there exists some variable assignment satisfying all constraints.

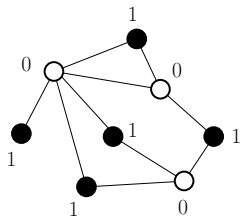


- Proper k -colorings of random graph (of average degree d).
- Independent sets (hardcore model) of a random graph.
- Random k -satisfiability.
- Using theory of spin glasses (random instances of satisfiability)...

Clustering of the solution space in random CSPs



- Replica symmetric (decay of correlations) vs. replica symmetry breaking (long range correlations).
- [MM '06] conjectured that 1 step replica symmetry breaking occurs exactly when the K-S bound is not tight.
- For $d < d_c(k)$ almost all solutions are contained in a giant cluster.
- For $d > d_c(k)$, there are exponentially many clusters.
[Mezard-Mora-Zecchina '05], [Achlioptas-Coja-Oghlan '08]
- [Mezard-Montanari '06] conjectured that clustering threshold coincides with tree reconstruction. Shown asymptotically in [Montanari-Restrepo-Tetali '09].



- Probability distribution over independent sets (ISs), fugacity $\lambda > 0$,

$$\mathbb{P}(\sigma) = \frac{1}{Z} \lambda^{|\sigma|} \mathbf{1}(\sigma \text{ is an independent set})$$

- Partition function Z

$$Z = \sum_{\sigma \in I(G)} \lambda^{|\sigma|}$$

Reconstruction for hardcore measure on the tree T_d

- Broadcast process for the hardcore measure on T_d

$$M = \begin{pmatrix} p_{11} & p_{10} \\ p_{01} & p_{00} \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ \frac{\alpha}{1-\alpha} & \frac{1-2\alpha}{1-\alpha} \end{pmatrix}$$

- Density of independent sets $\alpha = \alpha(\lambda, d)$ where

$$\lambda = \frac{\alpha}{1-\alpha} \left(\frac{1-\alpha}{1-2\alpha} \right)^d.$$

- Reconstruction solvable for λ if $\mathbb{P}(\sigma_r = 1 | \sigma_n) \not\rightarrow \alpha(\lambda, d)$.

- [Brightwell-Winkler '04] Reconstruction solvability for

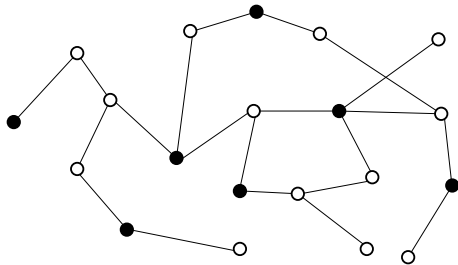
$$\alpha \geq \frac{\ln d + \ln \ln d + 1 + o(1)}{d}, \quad \lambda \geq (e + o(1))(\ln d)^2.$$

- [Martin '03] Non-solvable for $\lambda \leq e - 1$.

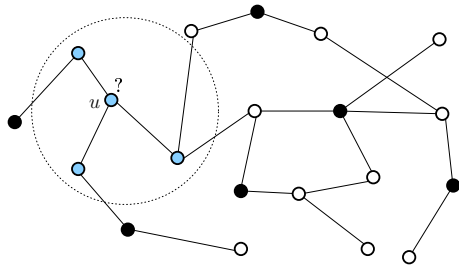
- [B-Sly-Tetali '10] Non-solvable for $\lambda \leq \frac{(\ln 2 - o_d(1)) \ln(d)}{2 \ln \ln d}$,

$$\alpha \leq \frac{\ln d + \ln \ln d - \ln \ln \ln d - \ln 2 + \ln \ln 2 - o(1)}{d},$$

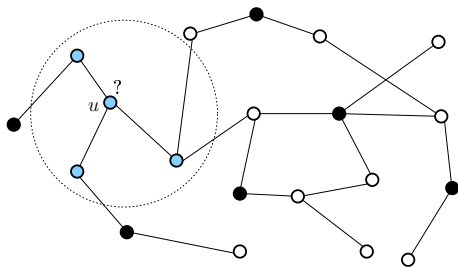
Reconstruction in graphs



Reconstruction in graphs



Reconstruction in graphs



- $\{G_n\}$ a family of random graphs, $|V(G_n)| \rightarrow \infty$.
- $\{G_n\}$ has non-reconstruction if for uniform $u \in V(G_n)$.

$$\lim_{L \rightarrow \infty} \limsup_n \mathbb{E} \left| \mathbb{P}(\sigma_u = 1 | \sigma(\partial B_u(L)), u) - \alpha(\lambda, d) \right| = 0$$

Reconstruction in graphs through local weak convergence

Theorem (B-Sly-Tetali '16)

For $\lambda < \lambda_c$, and large enough d , the d -regular random graph has non-reconstruction iff the d -regular tree has non-reconstruction.

$$\mathbb{P}_{T_d}(\sigma_x = 1 | \sigma_{L_\ell}) \xrightarrow{\mathbb{P}} \alpha(\lambda, d) \text{ as } \ell \rightarrow \infty.$$

\Leftrightarrow

$$\lim_{\ell \rightarrow \infty} \limsup_n \mathbb{E} |\mathbb{P}_{G_n}(\sigma_u = 1 | \sigma(\partial B_u(\ell))) - \alpha(\lambda, d)| = 0$$

where λ_c is the fugacity corresponding to

$$\alpha_c = \frac{2 \ln d - (3 + o(1)) \ln \ln d}{d}.$$

Convergence in probability locally

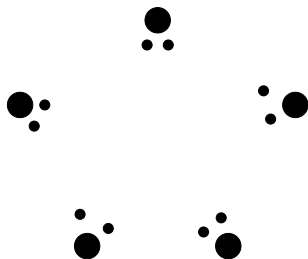
[Montanari-Mossel-Sly '12] Convergence in probability locally:

- For all $r, \varepsilon > 0$

$$\lim_{n \rightarrow \infty} \mathbb{P}_{U_n} (d_{tv}(\mathbb{P}_{G_n}(\sigma(B_r(u)) \in \cdot), \mathbb{P}_{T_d}(\sigma(B_r(x)) \in \cdot)) > \varepsilon) = 0$$

- Idea: use a type of second moment argument to show convergence in probability locally of hardcore measure to tree measure.

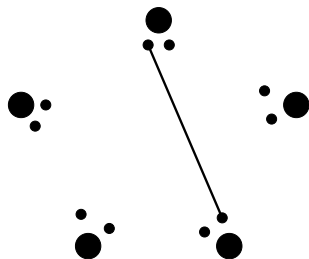
The configuration model



- Analyze the configuration model - d -regular multigraphs on n -vertices $\mathcal{C}(n, d)$.

$$\mathbb{P}(\text{Simple}) = (1 + o(1)) \exp\left(\frac{1 - d^2}{4}\right)$$

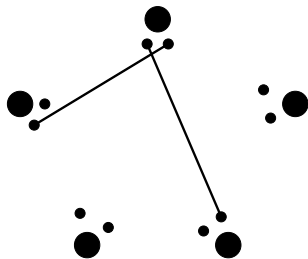
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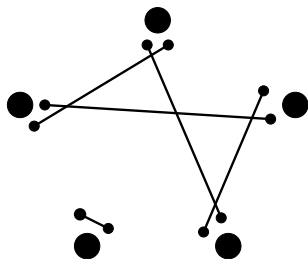
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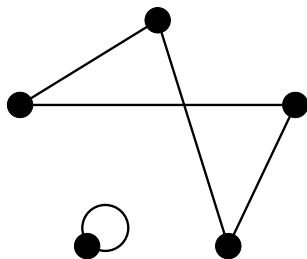
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The partition function of the hardcore model on $\mathcal{C}(n, d)$

$$Z_{G,\alpha} := \sum_{\sigma:|\sigma|=\alpha n} \lambda^{\alpha n}, \quad Z_G := \sum_{\alpha} Z_{G,\alpha}.$$

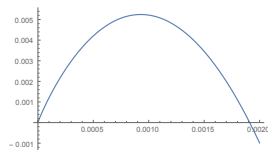


Figure: $\Phi(\alpha)$ vs. α .

$$\mathbb{E}(Z_{G,\alpha}) = \binom{n}{\alpha n} \lambda^{\alpha n} \prod_{i=0}^{\alpha nd-1} \frac{(1-\alpha)nd-i}{nd-1-2i} = \tilde{\Theta}(1) \exp(n\Phi(\alpha)).$$

$\Phi(\alpha)$ maximized at α^* which solves $\lambda = \frac{\alpha}{1-\alpha} \left(\frac{1-\alpha}{1-2\alpha} \right)^d$.

$$\alpha^*(\lambda, d) = (1 + o_d(1)) \frac{\ln(\lambda d / \ln(\lambda d))}{d}.$$

The second moment of $Z_{G,\alpha}$

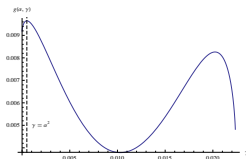
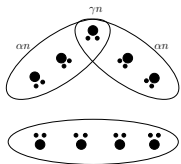
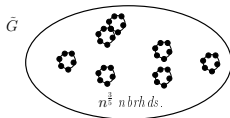


Figure: Projection $g(\alpha, \gamma)$ of f onto the maximizer $\bar{\varepsilon}(\alpha, \gamma)$.

Main facts/results: $G \sim \mathcal{C}(n, d)$

- $\mathbb{E}((Z_{G,\alpha})^2) = \sum_{\gamma, \varepsilon} \mathbb{E}(Z_{G,\alpha,\gamma,\varepsilon}^{(2)}) = \exp(nf(\alpha, \gamma, \varepsilon) + O(\ln(n)))$
- For any $0 \leq \alpha \leq \frac{1}{2}$, $2\Phi(\alpha) = f(\alpha, \alpha^2, \alpha(1 - 2\alpha))$.
- Let $\lambda < \lambda_c$. The function f is maximized at $(\alpha^*, (\alpha^*)^2, \alpha^*(1 - 2\alpha^*))$ and decays quadratically about this point.
- $\mathbb{E}((Z_G)^2) = \tilde{\Theta}(1)(\mathbb{E}(Z_G))^2$.

Relating graph and tree measures: punctured graph \tilde{G}



\tilde{G} : Choose $n^{3/5}$ vertices at random and delete their local neighborhood of radius r .

Study Z_G conditioned on the boundaries of local neighborhoods B .

Main results:

- Whp $O(n^{3/5})$ vertices of B have degree $d - 1$ and $O(n^{1/5})$ have degree $d - 2$.
- $\mathbb{E}(Z_{\tilde{G},\sigma})$ is proportional to a product measure on B .
- Second moment calculation for \tilde{G} . For each $\sigma \in \{0, 1\}^B$

$$\mathbb{E}((Z_{G,\sigma})^2) = \exp\left(O(n^{1/5})\right) \mathbb{E}(Z_{\tilde{G},\sigma})^2 \quad (1)$$

A set of bad boundaries \mathcal{B}

Local weak convergence: Let $G_n \sim \mathcal{C}(n, d)$. Then,

$$\frac{|\{i : \|\mathbb{P}_{G_n}(\partial B_r(s_i) \in \cdot), \mathbb{P}_{T_d}(\partial B_r(x) \in \cdot)\|_{tv} > \varepsilon\}|}{n^{3/5}} \xrightarrow{\mathbb{P}} 0$$

If not, there is a set \mathcal{B} of boundary configurations on which graph and tree measure differ.

- By (1) and Azuma's ineq, \mathcal{B} has large stationary prob. :

$$\mathbb{P}\left(\sum_{\sigma \in \mathcal{B}} Z_{G,\sigma} \geq \exp\left(-O(n^{\frac{1}{2} + \delta(\varepsilon)})\right) \mathbb{E}(Z_G)\right) \geq 1 - 2 \exp\left(-n^{2\delta(\varepsilon)}\right).$$

- $\mathbb{E}(Z_G)$ and $\mathbb{E}\left(\sum_{\sigma \in \mathcal{B}} Z_{G,\sigma}\right)$ related by tree measure $\mathbb{P}_{T_d}(\mathcal{B})$.
- On these events, the part. fn. is much larger than expected

$$\sum_{\sigma \in \mathcal{B}} Z_{G,\sigma} \geq \exp\left(cn^{\frac{3}{5}}\right) \mathbb{E}\left(\sum_{\sigma \in \mathcal{B}} Z_{G,\sigma}\right).$$

Second moment implies \mathcal{B} is unlikely

$$\begin{aligned} & \mathbb{P} \left(\sum_{\sigma \in \mathcal{B}} Z_{G,\sigma} \geq \exp \left(-O(n^{\frac{1}{2}+\delta}) \right) \mathbb{E}(Z_G) \right) \\ & \leq \frac{\mathbb{E} \left(\sum_{\sigma \in \mathcal{B}} Z_{G,\sigma} \right)}{\exp \left(-O(n^{\frac{1}{2}+\delta}) \right) \mathbb{E}(Z_G)} \quad (\text{Markov's ineq.}) \\ & \leq \frac{\exp \left(-cn^{\frac{3}{5}} \right)}{\exp \left(-O(n^{\frac{1}{2}+\delta}) \right)} \leq \exp \left(-cn^{\frac{3}{5}} \right) \end{aligned}$$

by choosing δ such that $\frac{3}{5} > \frac{1}{2} + \delta$.

Phase transitions for the hardcore measure

- [Barbier-Krzakala-Zdeborová-Zhang '13] Several earlier works showed the hardcore model undergoes a continuous "full replica symmetry breaking" transition for graphs of small degree.
- [Coja-Oghlan-Efthymiou '11] Clustering for hardcore measure in Erdős-Rényi graphs beyond $(1 + o_d(1)) \log(d)/d$. [Rahman-Virag '14, Gamarnik-Sudan '14] random regular graphs.
- [BKZZ13] use cavity methods to show that for degree $d > 16$, there is a discontinuous phase transition while for $d < 16$, it is continuous.
- [BKZZ13] "1 RSB approach": condensation transition exists and is at

$$\alpha_c \approx \frac{2(\log(d) + \log \log(d) + 1 - \log(2))}{d} + o(1/d).$$

Thank you for your attention!