

# Random cluster model on locally tree-like regular graphs

## Free energy and typical landscape

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Partition Functions, 28.05.2025.

Ongoing work with Ferenc Bencs and Péter Csikvári.

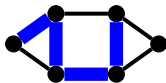
- 1 Definitions and setup
- 2 Results
- 3 Idea of the proof

# Random cluster model

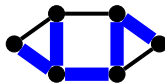
Let  $p \in (0, 1)$  and  $q > 0$  be fixed. The probability that  $A \subseteq E(G)$  is chosen is

$$\mathbb{P}[A] \propto p^{|A|} (1-p)^{|E(G) \setminus A|} q^{k(A)},$$

where  $k(A)$  denotes the number of connected components of  $(V, A)$ .



(a)  $\mathbb{P}[A_1] \propto p^4 (1-p)^4 q^2$



(b)  $\mathbb{P}[A_2] \propto p^5 (1-p)^3 q$

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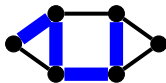
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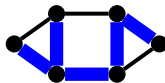
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For a graph  $G$  we denote the partition function of the random cluster model as

$$Z_G(q, w) = \sum_{A \subseteq E(G)} q^{k(A)} w^{|A|}.$$

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Examples:

- $G_n$  is a random  $d$ -regular graph on  $n$  vertices
- $G_{n+1}$  is a random double lift of  $G_n$  where  $G_0$  is any  $d$ -regular graph.

## Question

What is the free energy on an essentially large girth sequence of  $d$ -regular graphs:

$$\lim_{n \rightarrow \infty} \frac{1}{v(G_n)} \log Z_{G_n}(q, w)?$$



# Main questions

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## Question

What is local weak limit of this the random cluster model on  $G_n$ ? How does a typical configuration look like?

# Previous results

- Dembo, Montanari / Montanari, Mosser, Sly:  
Ising model ( $q = 2$ )
- Dembo, Montanari, Sun:  
Potts model (positive integer  $q$ ), except an interval  $(w_0, w_1)$
- Dembo, Montanari, Sly and Sun:  
Potts model (positive integer  $q$ ) and even  $d$
- Helmuth, Jenssen and Perkins:  
random cluster model with large  $q$  and assuming some expansion property of  $(G_n)_n$
- Basak, Dembo and Sly:  
local structure of the Potts model (positive integer  $q$ ) for all  $d$

## Theorem (BBCs '22)

*If  $(G_n)_n$  is an essentially large girth sequence of  $d$ -regular graphs, then the limit*

$$\lim_{n \rightarrow \infty} \frac{1}{v(G_n)} \ln Z_{G_n}(q, w) = \ln \Phi_{d,q,w}$$

*exists for  $q \geq 2$  and  $w \geq 0$ . The quantity  $\Phi_{d,q,w}$  can be computed as follows. Let*

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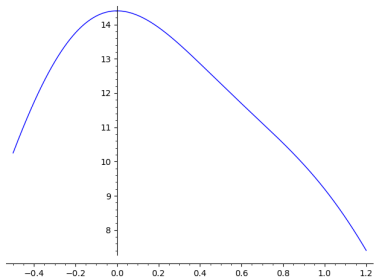
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$$\begin{aligned} \Phi_{d,q,w}(t) := & \left( \sqrt{1 + \frac{w}{q}} \cos(t) + \sqrt{\frac{(q-1)w}{q}} \sin(t) \right)^d \\ & + (q-1) \left( \sqrt{1 + \frac{w}{q}} \cos(t) - \sqrt{\frac{w}{q(q-1)}} \sin(t) \right)^d, \end{aligned}$$

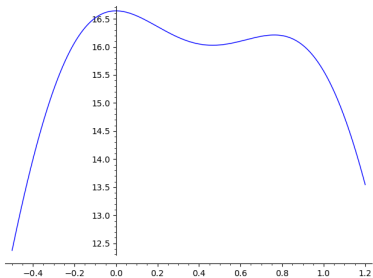
then

$$\Phi_{d,q,w} := \max_{t \in [-\pi, \pi]} \Phi_{d,q,w}(t).$$

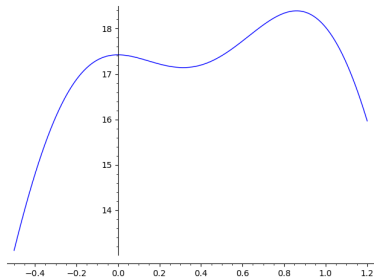
# Examples



$$q = 4, w = 2, d = 4$$



$$q = 4, w = 2.9, d = 4$$



$$q = 4, w = 3.2, d = 4$$

## Theorem (BBCs '22)

*Let  $q > 2$  and*

$$w_c := \frac{q-2}{(q-1)^{1-2/d} - 1} - 1.$$

*If  $0 \leq w \leq w_c$ , then  $\Phi_{d,q,w} = q \left(1 + \frac{w}{q}\right)^{d/2}$ .*

*If  $w > w_c$ , then  $\Phi_{d,q,w} > q \left(1 + \frac{w}{q}\right)^{d/2}$ .*

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*The model undergoes a first-order phase transition.*

## Theorem (BBCs '25+)

Let  $q > 2$ .

- If  $w < w_c$ , then the "typical" local structure of random cluster model on  $G_n$  "looks like" the free boundary condition Gibbs measure on  $T_d$ .
- If  $w > w_c$ , then the "typical" local structure of random cluster model on  $G_n$  "looks like" the wired boundary condition Gibbs measure on  $T_d$ .



# Local structure

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## Definition

For a configuration  $\sigma$  on  $G_n$  let  $S_r\sigma$  be the statistics of the  $r$ -balls.

## Theorem

For all  $\varepsilon > 0$  and  $r$  there exists  $c > 0$  such that

$$\mathbb{P}_{\sigma_n} [d_{TV} (S_r\sigma_n, \mu|_{B(o,r)}) \geq \varepsilon] < e^{-c\nu(G_n)}.$$

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- Ising model is well-understood.

## Ising model

Let  $\beta > 0, h \in \mathbb{R}$ . For a  $\sigma : V \rightarrow \{+, -\}$

$$\mathbb{P}[\sigma] = \frac{1}{Z_G(\beta, h)} e^{\beta \sum_{(u,v) \in E} \sigma_u \sigma_v + h \sum_{v \in V} \sigma_v},$$

where

$$Z_G(\beta, h) = \sum_{\sigma: V \rightarrow \{+, -\}} e^{\beta \sum_{(u,v) \in E} \sigma_u \sigma_v + h \sum_{v \in V} \sigma_v}.$$

## Spin models

Let  $N \in M^{r \times r}(\mathbb{R}_{\geq 0})$  symmetric and  $\nu \in \mathbb{R}_{\geq 0}^r$ . Then for a  $\sigma : V \rightarrow [r]$

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## Ising as a spin model

$$N_{\text{Ising}} = \begin{pmatrix} e^{\beta} & e^{-\beta} \\ e^{-\beta} & e^{\beta} \end{pmatrix} \text{ and } \nu_{\text{Ising}} = \begin{pmatrix} e^h \\ e^{-h} \end{pmatrix}.$$

# The approximating spin model

## 2-spin approximation

Let

$$N = \begin{pmatrix} 1+w & 1 \\ 1 & 1+\frac{w}{q-1} \end{pmatrix} \text{ and } \nu = \begin{pmatrix} 1 \\ q-1 \end{pmatrix}.$$



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## In the $d$ -regular case

$$\begin{pmatrix} 1+w & 1 \\ 1 & 1+\frac{w}{q-1} \end{pmatrix}, \begin{pmatrix} 1 \\ q-1 \end{pmatrix} \sim \begin{pmatrix} 1+w & a \\ a & \left(1+\frac{w}{q-1}\right)a^2 \end{pmatrix}, \begin{pmatrix} 1 \\ \frac{q-1}{a^d} \end{pmatrix}$$

# Colored random cluster

## Coloring the random cluster model

- 1 Sample from the random cluster,
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## Weight

$$w_{cRC}(A, \sigma) = \prod_{C \in \mathcal{C}_{cyclic}(A)} w^{e(C)} \cdot q \prod_{C \in \mathcal{C}_{tree}(A)} w^{e(C)} \cdot 1 \prod_{C \in \mathcal{C}_{tree}(A)} w^{e(C)} \cdot (q-1).$$

## Remark

$$w_{RC}(A) = \sum_{\sigma} w_{cRC}(A, \sigma).$$

# Percolated Ising model

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## Percolating the 2-spin model

- 1 Sample from the 2-spin model; percolate:

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## Percolating the 2-spin model

- 1 Sample from the 2-spin model; percolate:
- 2 blue-red edge: remove;
- 3 red-red edge: keep with probability  $\frac{w}{1+w}$ ,
- 4 blue-blue edge: keep with probability  $\frac{\frac{w}{q-1}}{1+\frac{w}{q-1}} = \frac{w}{w+q-1}$  (active edges).

## Weight

$$w_{pIs}(A, \sigma) = \prod_{C \in \mathcal{C}(A)} w^{e(C)} \cdot 1^{v(C)} \prod_{C \in \mathcal{C}(A)} \left( \frac{w}{q-1} \right)^{e(C)} \cdot (q-1)^{v(C)}$$

# Modified Edwards–Sokal coupling

## Weights

$C$	$\sigma(C)$	in percolated Ising	in colored RC
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cyclic	-	$w^{e(C)}(q-1)^{v(C)-e(C)}$	0

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## Subexponentially close weights

$$w_{pIs}(A, \sigma) \leq w_{cRC}(A, \sigma) \leq w_{pIs}(A, \sigma) \cdot q^{\frac{v(G_n)}{g}}$$

## Theorem (BBCs '22)

$$\frac{1}{v(G_n)} \ln Z_{Is}(q, w) - \frac{1}{v(G_n)} \ln Z_{RC}(q, w) \rightarrow 0.$$

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## Theorem (BBCs '25+)

$$\mathbb{P}_{pIS}(A_n) < e^{-cv(G_n)} \implies \mathbb{P}_{cRC}(A_n) < e^{-c'v(G_n)}.$$

## Lemma

For all  $\varepsilon > 0$  and  $r$  there exists  $c > 0$  such that

$$\mathbb{P}_{\sigma_n \sim pIS} \left[ d_{TV} \left( S_r \sigma_n, \mu|_{B(o,r)} \right) \geq \varepsilon \right] < e^{-cv(G_n)}.$$

THANKS FOR YOUR ATTENTION!