

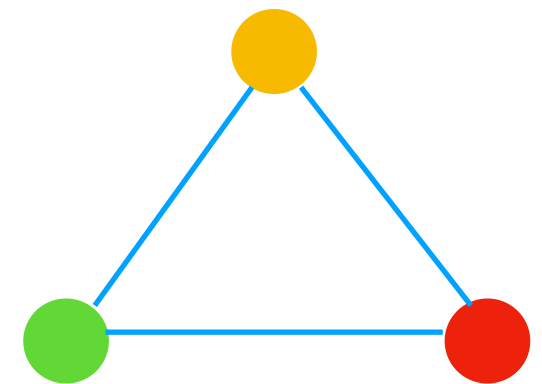
A Deterministic Algorithm for Counting Colorings with 2Δ Colors

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Approximate counting

- Consider proper q -colorings of a graph G
- Let $\Delta := \Delta(G)$ be the maximum degree
- If $q \geq \Delta + 1$:
 - G has a q -coloring, which can be found efficiently
 - How many q -colorings does G have?



#P-complete, approximation?

Approximate counting

- Compute $(1 \pm \varepsilon) \cdot (\#q\text{-colorings})$

- Equivalent to

- **(Approximate) sampling**

Sample a q -coloring uniformly at random?

- **Approximate inference**

Given the color of a vertex u ,

what can you infer about the color of another vertex v ?

- **Approximate root-finding**

- ...

Our main result

Our algorithmic result:

For any integer $q \geq 2\Delta$, and bounded-degree graph G , there exists a **deterministic FPTAS**, which outputs \hat{Z} s.t.

$$\hat{Z} \in (1 \pm \varepsilon)(\#q\text{-colorings}(G))$$

in time $\text{poly}(|G|, 1/\varepsilon)$ for G

Triangle-free graphs:

For any integer $q \geq 1.7633\Delta + \beta$, and triangle-free graph G of bounded degree, the above **FPTAS** also works

Dobrushin condition

[Gamarnik, Katz and Mishra] 1.7633 is closely related to strong spatial mixing, which is the unique positive solution of the equation $xe^{-1/x} = 1$

Prior works (incomplete list):

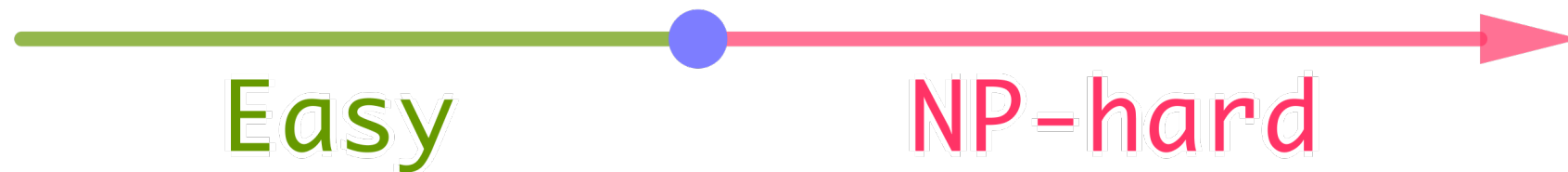
- **Randomized** MCMC for $q \geq 2\Delta + 1$ [Jerrum]
- **Randomized** MCMC for $q \geq (\frac{11}{6} - \epsilon)\Delta$ [Chen et al.] [Vigoda]
- **Deterministic** algorithm for $q \geq 2.58\Delta + 1$ [Lu and Yin]
- **Deterministic** algorithm for $q = 4, \Delta = 3$ [Lu et al.]

Computational phase transition

Our work: exploit a formal connection between algorithms and phase transition.

Crucially we exploit both directions

- Many problems have a “computational phase transition” (roughly):

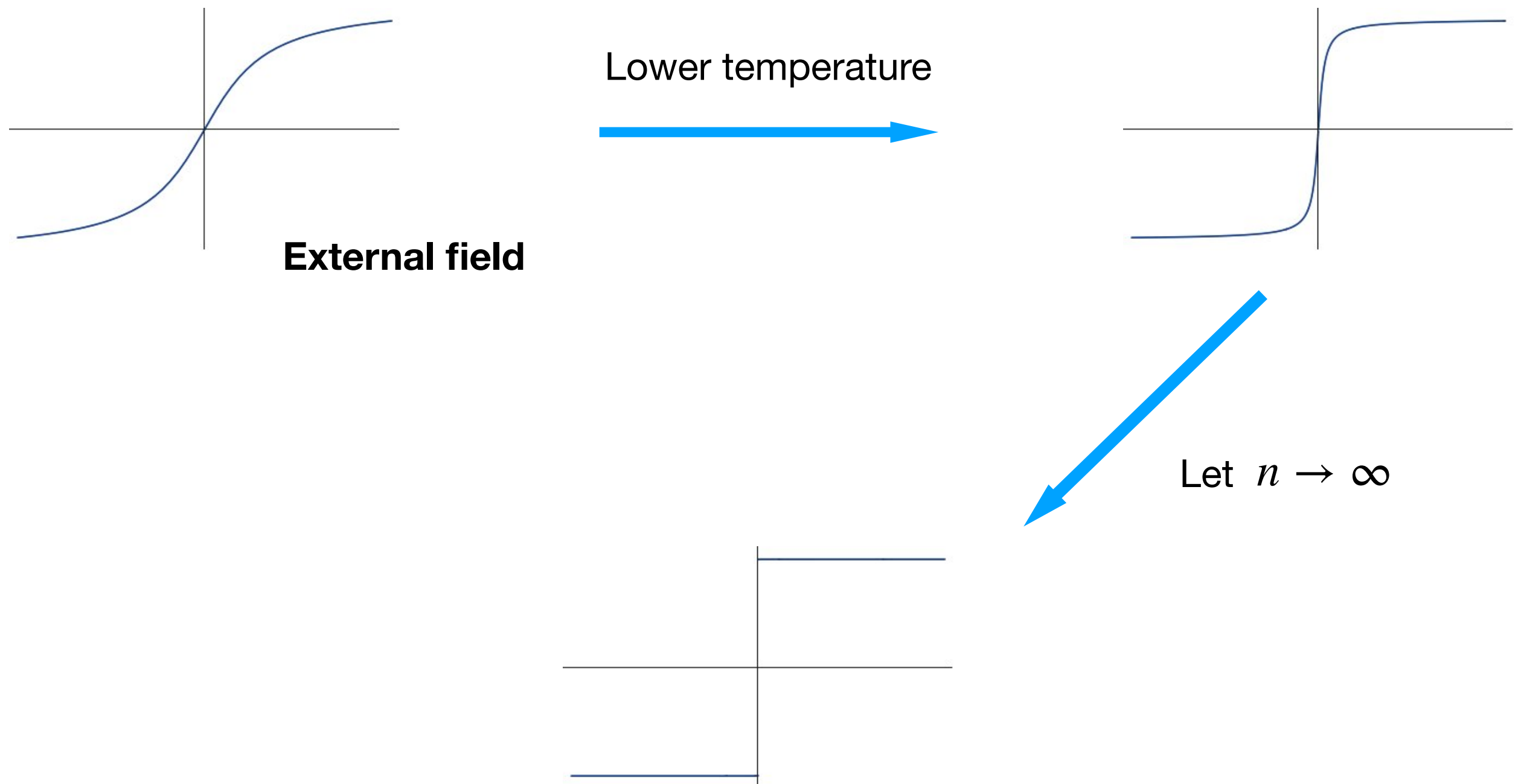


- In some cases, known to coincide with statistical physics phase transition:
 - Hardcore model
 - Anti-ferromagnetic Ising/2-spin model
 -

Phase transitions \approx Discontinuity

In statistical physics

Observable, e.g., mean magnetization \approx average number of +-spin vertices

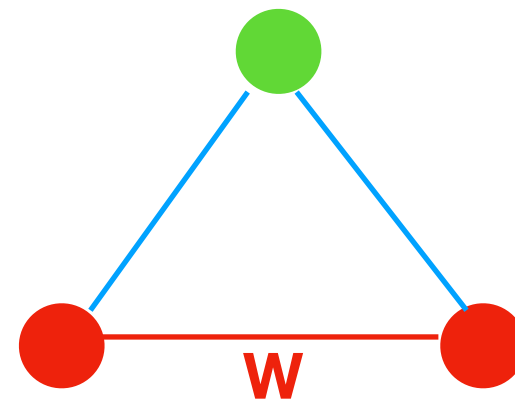


Approximate counting

The Potts model

- A graph $G=(V,E)$
- Configuration $\sigma : V \rightarrow [q]$
- The partition function

$$\begin{aligned} Z_G(w) &:= \sum_{\sigma: V \rightarrow [q]} w^{(\# \text{monochromatic edges})} \\ &= \sum_{\sigma: V \rightarrow [q]} w^{|E_=(\sigma)|}. \end{aligned}$$



Example:

- $Z_G(0) = (\# \text{ valid } q\text{-colorings of } G)$
- $q=2$: Ising model

- Gibbs distribution:

$$\Pr[\sigma] = \frac{1}{Z_G(w)} \cdot w^{|E_=(\sigma)|}$$

Goal: compute $(1 \pm \varepsilon) \cdot Z_G(w)$.

Phase transitions formally:

Geometry of polynomials

- Lee-Yang theory:

Phase transition \approx complex zeros of Z

$\log Z$

- Recall the partition function: $Z_G(w) := \sum_{\sigma: V \rightarrow [q]} w^{|E_=(\sigma)|}$

- Observable:

$$\begin{aligned}\mathbb{E}_\sigma |E_=(\sigma)| &= \sum_{\sigma: V \rightarrow [q]} |E_=(\sigma)| \cdot \Pr[\sigma] \\ &= w \cdot \frac{\partial \log Z_G(w)}{\partial w}\end{aligned}$$

- Analyticity of $\log Z \approx$ Continuity of observables

Lack of phase transition \approx Lack of complex zeros

Algorithms from Phase transitions: Barvinok's interpolation

Recall the partition function: $Z_G(w) := \sum_{\sigma: V \rightarrow [q]} w^{|E_=(\sigma)|}$

[Barvinok, Barvinok and Soberon]

- Consider the Taylor expansion of $\log Z$
- In a zero free region, $\log Z$ can be approximated to $\pm \epsilon$ by its k -th order Taylor series for $k = O(\log(n/\epsilon))$
- $\log Z_G(w) \pm \epsilon \iff (1 \pm \epsilon) \cdot Z_G(w)$
- k -th order Taylor series is determined by the first $k+1$ coefficients of Z
- Naively computing the first $k+1$ coefficients of Z takes time $O(n^k)$
- \implies **Quasi-polynomial** time algorithm for $k = O(\log(n/\epsilon))$
- Exploiting the combinatorial structure speeds up to $O(n(e\Delta)^k)$

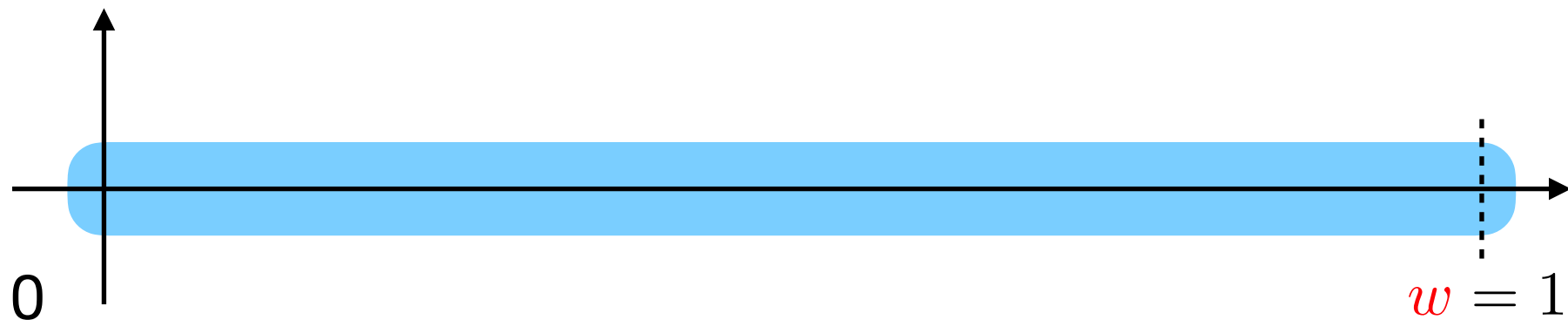
Fisher zeros of the Potts model

Recall the partition function:

$$Z_G(\textcolor{red}{w}) := \sum_{\sigma: V \rightarrow [\textcolor{blue}{q}] } \textcolor{red}{w}^{|E_=(\sigma)|}$$

Our zero-freeness result:

Fix any integer q such that $q \geq 2\Delta$. Then there exists a constant τ_Δ such that $Z_G(\textcolor{red}{w}) \neq 0$ when $\textcolor{red}{w} \in B([0, 1], \tau_\Delta)$



Prior to our work:

- $q \geq e\Delta + 1$ [Bencs. et. al.]: similar region
- $q \geq 7.964\Delta$ [Sokal] or $q \geq 6.907\Delta$ [Fernández and Procacci]: entire unit disk

Potts model (triangle-free)

$$Z_G(\boldsymbol{w}) := \sum_{\sigma: V \rightarrow [q]} \boldsymbol{w}^{|E_=(\sigma)|}$$

Triangle-free graphs:

Fix any integer q such that: $q \geq 1.7633\Delta + \beta$, there exists a constant τ_Δ such that


$Z_G(\boldsymbol{w}) \neq 0$ if G is triangle-free and $\boldsymbol{w} \in B([0, 1], \tau_\Delta)$



\implies **FPTAS** for bounded degree graphs

Phase transitions from analysis of algorithms

Zero-freeness using algorithmic ideas


$$Z_G(w) := \sum_{\sigma: V \rightarrow [q]} w^{|E_{=}(\sigma)|}$$

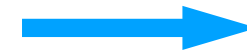
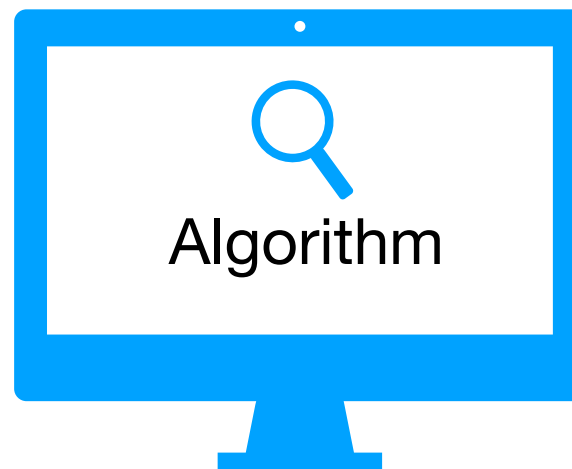
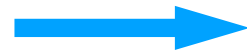
If G has a valid q -coloring, then

$$Z_G(w) \geq 1$$

It suffices to prove that

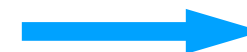
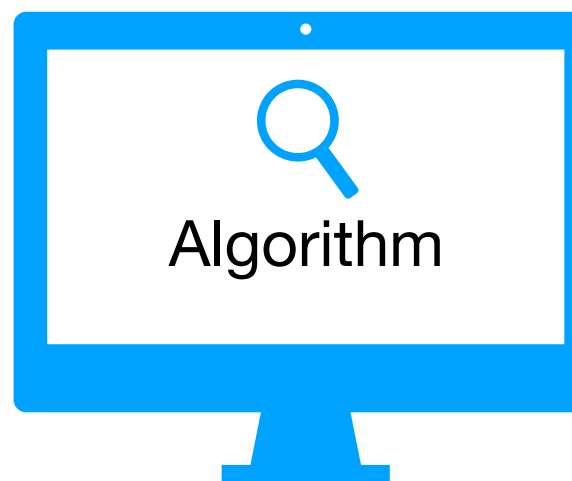
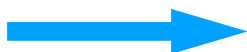
$$\left| \frac{Z_G(\tilde{w})}{Z_G(w)} \right| > 0$$

$$w \in [0, 1]$$



$$Z_G(w)$$

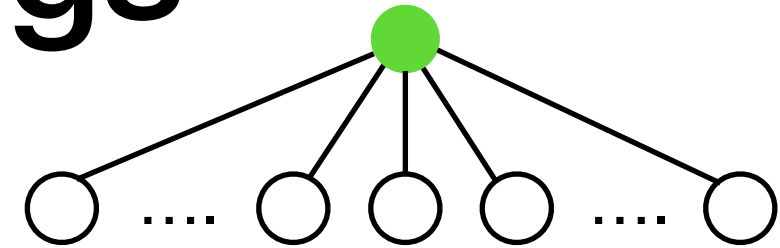
$$\tilde{w} \approx w, \tilde{w} \in \mathbb{C}$$



$$Z_G(\tilde{w})$$

Ratios and Pinning

Fix a vertex v



$Z_{G,v}^{(i)}(\omega)$: colorings in which vertex v receives color i

$$Z_G(\omega) = \sum_{i \in L(v)} Z_{G,v}^{(i)}(\omega)$$

$$= Z_{G,v}^{(j)}(\omega) \cdot \sum_{i \in L(v)} \frac{Z_{G,v}^{(i)}(\omega)}{Z_{G,v}^{(j)}(\omega)}$$

$$Z_G(\tilde{\omega}) = \sum_{i \in L(v)} Z_{G,v}^{(i)}(\tilde{\omega})$$

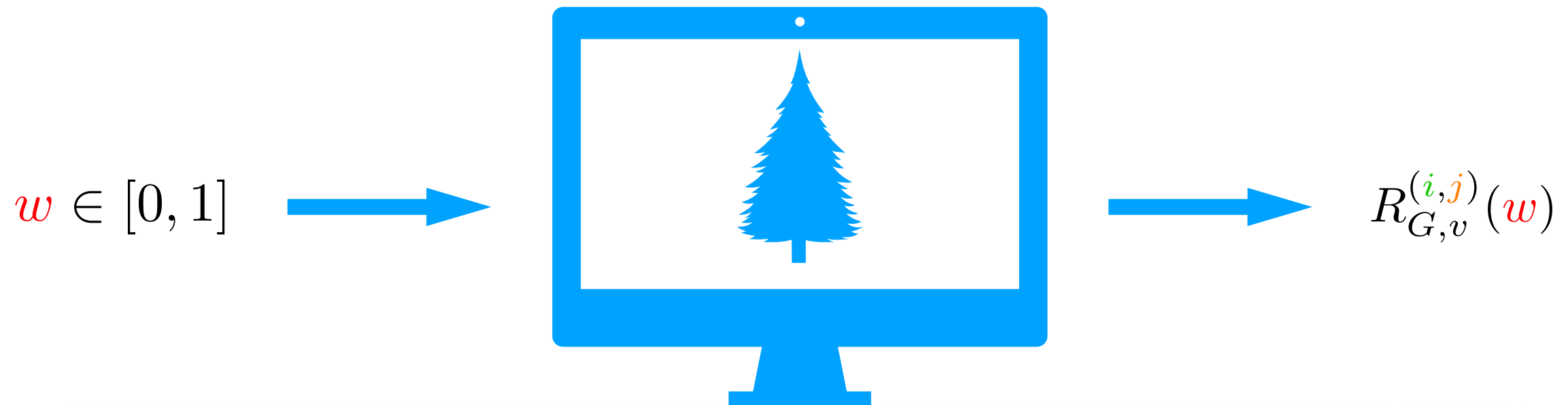
$$= Z_{G,v}^{(j)}(\tilde{\omega}) \cdot \sum_{i \in L(v)} \frac{Z_{G,v}^{(i)}(\tilde{\omega})}{Z_{G,v}^{(j)}(\tilde{\omega})}$$

Induction

Consider $R_{G,v}^{(i,j)}(\omega) := \frac{Z_{G,v}^{(i)}(\omega)}{Z_{G,v}^{(j)}(\omega)}$ inductively

We show that $R_{G,v}^{(i,j)}(\tilde{\omega}) \approx R_{G,v}^{(i,j)}(\omega)$

Tree recurrence



At every vertex of the tree, the two computations remain close



We treat the tree recurrence as a complex dynamical system:

- Not every orbit is well-behaved (unless $q \gg 2\Delta$): main barrier
- Insight: There exists a “*high entropy*” (nice) orbit, which is well-behaved

High entropy condition

Consider list-coloring

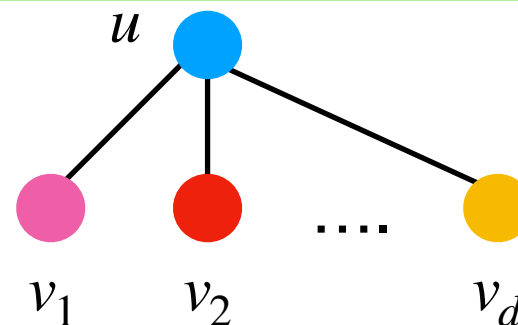
For any vertex u in G , any $w \in [0, 1]$ and any color i ,

$$\Pr_{G,w}[c(u) = i] \leq \frac{1}{\Delta + 2}$$

Lemma:

For any list-coloring instance such that $q \geq 2\Delta + 2$, or the graph is triangle-free and $q > 1.76\Delta + \beta$, then our “high entropy condition” holds

A more careful argument shows that $q \geq 2\Delta$ suffices



Conjecture:

For graphs of higher girth, $q \geq \alpha\Delta + \beta$ for a smaller α should suffice

Summary of the induction

$$Z_G(\textcolor{red}{w}) = \sum_{\textcolor{green}{i} \in L(v)} Z_{G,v}^{(\textcolor{green}{i})}(\textcolor{red}{w})$$

$$= Z_{G,v}^{(\textcolor{brown}{j})}(\textcolor{red}{w}) \cdot \sum_{\textcolor{green}{i} \in L(v)} \frac{Z_{G,v}^{(\textcolor{green}{i})}(\textcolor{red}{w})}{Z_{G,v}^{(\textcolor{brown}{j})}(\textcolor{red}{w})}$$

Inductively

$$\left| \frac{Z_{G,v}^{(\textcolor{brown}{j})}(\textcolor{red}{\tilde{w}})}{Z_{G,v}^{(\textcolor{brown}{j})}(\textcolor{red}{w})} \right| \geq e^{-\epsilon(n-1)}$$

$$Z_G(\textcolor{red}{\tilde{w}}) = \sum_{\textcolor{green}{i} \in L(v)} Z_{G,v}^{(\textcolor{green}{i})}(\textcolor{red}{\tilde{w}})$$

$$= Z_{G,v}^{(\textcolor{brown}{j})}(\textcolor{red}{\tilde{w}}) \cdot \sum_{\textcolor{green}{i} \in L(v)} \frac{Z_{G,v}^{(\textcolor{green}{i})}(\textcolor{red}{\tilde{w}})}{Z_{G,v}^{(\textcolor{brown}{j})}(\textcolor{red}{\tilde{w}})}$$

$$R_{G,v}^{(\textcolor{green}{i}, \textcolor{brown}{j})}(\textcolor{red}{\tilde{w}}) \approx e^\epsilon \cdot R_{G,v}^{(\textcolor{green}{i}, \textcolor{brown}{j})}(\textcolor{red}{w})$$

Recap

- One can exploit the *analysis of algorithms* (**tree recurrence**) to study *phase transitions* (prove zero-freeness)

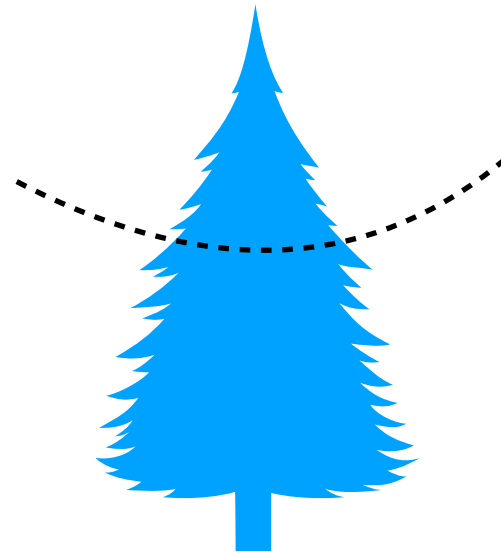
$$\frac{Z_{G,v}^{(i)}(w)}{Z_{G,v}^{(j)}(w)} = \frac{Z_{G,v}^{(i)}(\tilde{w})}{Z_{G,v}^{(j)}(\tilde{w})}$$

- Conversely, zero-freeness results can also be exploited algorithmically (via Barvinok's interpolation)

Discussion: Comparison to decay of correlations

Correlation decay approach

Truncating



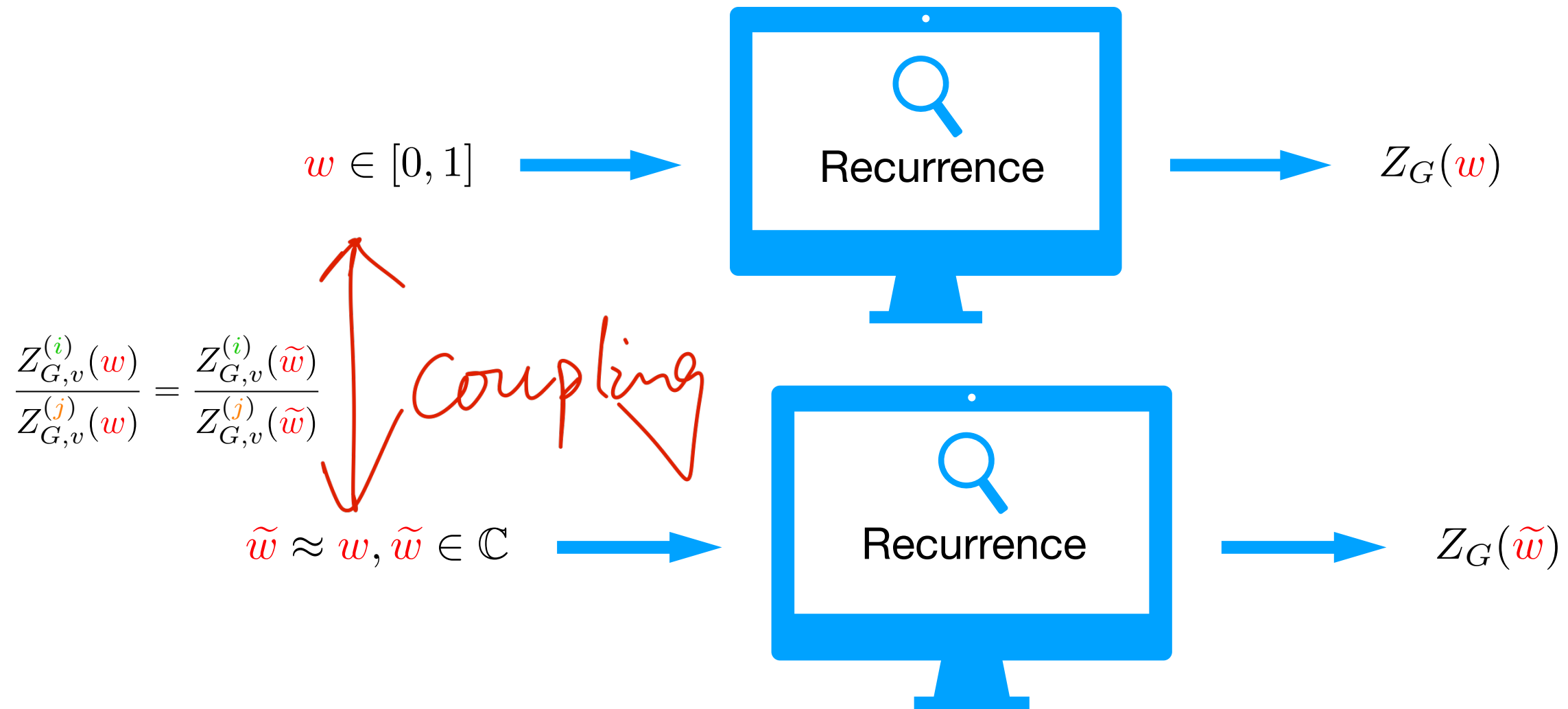
Guess the values

Tree recurrence

Issue: guessing the values **explicitly** can be hard

In proving zero-freeness, we only need that “good” values **exist**

Comparison to MCMC



Future direction:

- Generalize to more sophisticated coupling argument
- De-randomize MCMC

Thanks

Q & A