

Interpolation

Sunday, April 29, 2018 10:34 PM

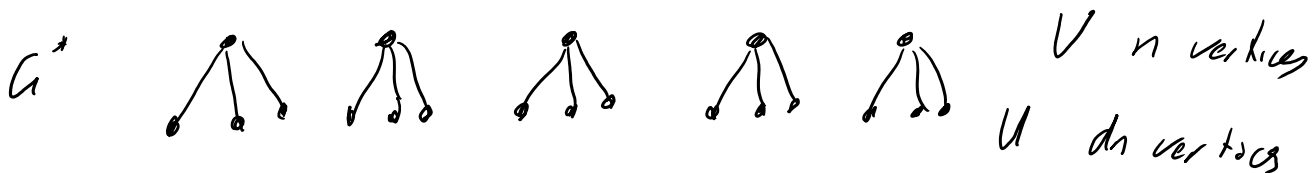
We want to move from G_N to G_{N+n} , how does the partition function change?



If we know $\hat{N}(\sigma_U)$ we could calculate

$$\frac{Z_{N+n}}{Z_N} = \frac{Z_{N+n}}{\hat{Z}} \cdot \frac{\hat{Z}}{Z_N}$$

Interpolation Method and Upper Bounds on free energy



Model:

$$P[\sigma] = \frac{1}{Z^*} \lambda^{\sum_U \sigma_U} \prod_{\substack{u,v \\ u \in U, v \in V}} (1 - (1-\delta)\sigma_u \sigma_v) \cdot \mu(\sigma_1, \dots, \sigma_n)$$

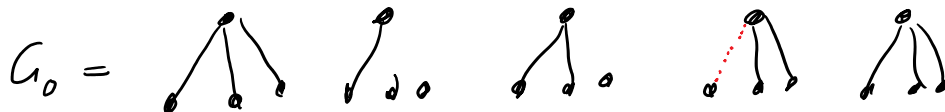
* absorb $\lambda \sum \sigma_n$ into $\pi m_u(\sigma_n)$.

↳ remove $n^{2/3}$ edges

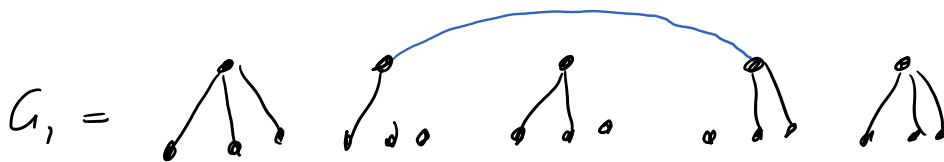


• For $r = 0, \dots, n - n^{2/3} - 1$

- Pick an edge (u, v) in G_r , $u \in U, v \in V$
Remove it to form G_r'

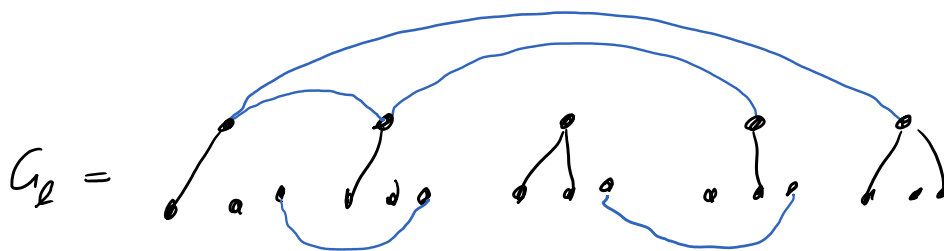


- With prob $\frac{1}{2}$, connect two half edges $v_1, v_2 \in V$.
- Otherwise connect two half edges $u_1, u_2 \in U$.

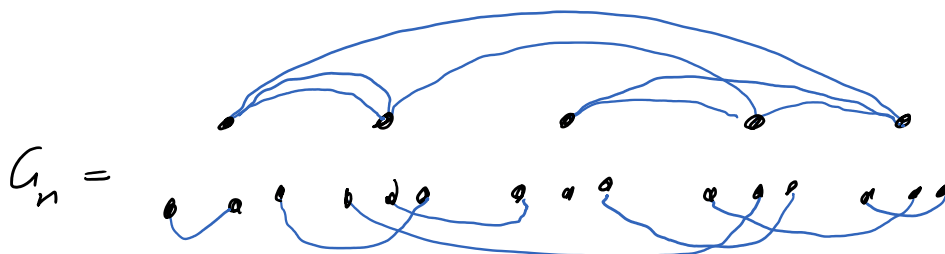


This is G_{r+1} .

G_n



- Form G_n by matching remaining half edges within U and within V .



$$\text{Let } \Delta_r = \mathbb{E} \log \frac{Z_{r+1}}{Z_r} = \mathbb{E} \log \frac{Z_{r+1}}{Z_r'} - \log \frac{Z_r}{Z_r'}$$

Let $u, u' \in U, v, v' \in V$ be random unmatched half edges

$$\mathbb{E} \log \frac{Z_r}{Z_r'} = \mathbb{E} \log \mathbb{E}[(1 - (1-\delta)\sigma_u \sigma_v) | G_{r-1}, u, v]$$

$$= -\mathbb{E} \sum_{p \geq 1} \frac{(1-\delta)^p}{p} \mathbb{E}[\sigma_u \sigma_v | G_{r-1}, u, v]^p$$

Let $\{\sigma^e\}_{e \geq 1}$ be IID copies of σ given G_{r-1} .

and $Z_u^p = \prod_{e=1}^p \sigma_u^e$.

$$\begin{aligned}
&= -\mathbb{E} \sum_{p \geq 1} \frac{(1-\delta)^p}{p} \mathbb{E} \left[\left(\prod_{e=1}^p \sigma_n^e \right) \cdot \left(\prod_{e=1}^p \sigma_u^e \right) \mid G_{r', u, v} \right] \\
&= -\mathbb{E} \sum_{p \geq 1} \frac{(1-\delta)^p}{p} \mathbb{E} \left[\tau_u^p \tau_v^p \mid G_{r', u, v} \right]
\end{aligned}$$

Now let $A_p = \mathbb{E}[\tau_u^p \mid G_{r', \tau}]$, $B_p = \mathbb{E}[\tau_v^p \mid G_{r', \tau}]$

$$= -\sum_{p \geq 1} \frac{(1-\delta)^p}{p} \mathbb{E} \left[\mathbb{E}[A_p B_p \mid G_{r', \tau}] \right]$$

$$\begin{aligned}
\mathbb{E} \log \frac{Z_{r+1}}{Z_r} &= \frac{1}{2} \mathbb{E} \log \mathbb{E} \left[(1 - (1-\delta)\sigma_n \sigma_{u'}) \mid G_{r', u, u'} \right] \\
&\quad + \frac{1}{2} \mathbb{E} \log \mathbb{E} \left[(1 - (1-\delta)\sigma_v \sigma_{v'}) \mid G_{r', v, v'} \right]
\end{aligned}$$

$$= -\mathbb{E} \sum_{p \geq 1} \frac{(1-\delta)^p}{p} \left(\mathbb{E}[\sigma_n \sigma_{u'} \mid G_{r', u, u'}]^p + \mathbb{E}[\sigma_v \sigma_{v'} \mid G_{r', v, v'}]^p \right) / 2$$

$$= -\sum_{p \geq 1} \mathbb{E} \mathbb{E} \left[\frac{\tau_u^{(p)} \tau_{u'}^{(p)}}{2} + \frac{\tau_v^{(p)} \tau_{v'}^{(p)}}{2} \mid G_{r', u, u', v, v'} \right]$$

$$= -\sum_{p \geq 1} \frac{(1-\delta)^p}{p} \mathbb{E} \left[\mathbb{E} \left[(A_p^2 + B_p^2) / 2 \mid G_{r', \tau} \right] \right]$$

So

$$\mathbb{E} \Delta_r = -\sum_{p \geq 1} \frac{(1-\delta)^p}{p} \mathbb{E} \left[\mathbb{E} \left[(A_p^2 + B_p^2) - A_p B_p \mid G_{r', \tau} \right] \right]$$

≤ 0 .

So $\mathbb{E} \log G_n \leq \mathbb{E} \log G^* + O(n^{-3/2})$.

- G_n is a random d -regular graph plus $\frac{d}{n}$ disconnected edges with external field m .
- G_n^* is n depth 1 trees with depth 1.

$$\text{Let } \Phi_V = \frac{1}{n} \log Z_{G_n^*} \quad \Phi_E = \frac{2}{dn} \log Z_{G_n \setminus \cup}$$

$$\Phi_{Res_n} = \frac{1}{n} \log Z_{G_n \setminus V}$$

We have shown that for any choice of μ ,

$$\mathbb{E} \Phi_V \geq \mathbb{E} \frac{d}{2} \Phi_E + \Phi_{Res_n} + O(n^{-1/3})$$

and so

$$\mathbb{E} \Phi_{Res_n} \leq \mathbb{E} \left(\Phi_V - \frac{d}{2} \Phi_E \right)$$

$$\Leftrightarrow \mathbb{E} \log Z_n \leq \mathbb{E} \log Z_{N+n} - \log Z_n$$

$$\Leftrightarrow \mathbb{E} \log Z_n + \log Z_N \leq \mathbb{E} \log Z_{N+n} + O(n^{-1/3})$$

Super additivity $\Rightarrow \Phi = \lim \frac{1}{n} \log Z_n$ exists!

Plugging in \mathcal{M}

"Replica Symmetric Bound"

If \mathcal{M} is an IID measure with marginals m

$$\mathbb{E} \Phi_{Res_n}$$

$$\leq \inf_m \mathbb{E} \log \left(\sum_{x_u \in (0,1)} \lambda^{x_u} \prod_{u < v} \sum_{x_u} m(x_u) \log(1 - (1-\delta)x_u x_v) \right) \\ - \frac{d}{2} \mathbb{E} \log \left[\sum_{x_u, x_{u'}} m(x_u) m(x_{u'}) \log(1 - (1-\delta)x_u x_{u'}) \right]$$

1-Step Replica Symmetry Breaking (RSB) Bound

A Poisson Dirichlet Process with parameter $\delta \in (0,1)$ are the points of a Poisson process with intensity $x^{-1-\delta}$, normalized to have sum 1.

Denote it $(a_i)_{i \geq 1}$. Let $\{z_i\}$ be IID, then

$$\mathbb{E} \log \sum a_i z_i = \log(\mathbb{E}(z_i^\delta)^{1/\delta}) + \log(\sum a_i) \\ = \frac{1}{\delta} \log \mathbb{E}(z_i^\delta).$$

P.f.: If Π is Poisson Point Process with density $x^{-1-\delta}$, $\tilde{\Pi} = \{x\}_x : x \in \Pi\}$, also a P.P.P. with intensity at x ,

$$\int f_3(u) \mathbb{E}[\# \text{pts in } [\frac{x}{u}, \frac{x}{u} + \frac{dx}{u}]] dx$$

$$\int f_3(u) \mathbb{E}[\#\text{pts in } L_{\bar{u}, \bar{u} + \frac{1}{u}}] du$$

$$= \int f_3(u) \left(\frac{x}{u}\right)^{-1-\delta} \frac{dx}{u} \cdot du$$

$$= \int x^{-1-\delta} dx \cdot \mathbb{E}[\zeta^\delta].$$

Let $\zeta'_x = \zeta_x / \mathbb{E}[\zeta^\delta]^{1/\delta}$, then

$$\tilde{\pi} = \{x \zeta'_x : x \in \pi\} \stackrel{d}{=} \pi.$$

$$\mathbb{E} \log \sum_{x \in \pi} x \zeta'_x = \mathbb{E} \log \sum_{x \in \pi} x \cdot \zeta_x / \mathbb{E}[\zeta^\delta]^{1/\delta}$$

$$= \mathbb{E} \log \sum_{x \in \pi} x \zeta_x - \log \mathbb{E}[\zeta^\delta]^{1/\delta}$$

$$= \mathbb{E} \log \sum_{x \in \pi} x$$

$$(a_i)_{i \geq 1} = \left(\frac{x_i}{\sum x_j} \right)_{i \geq 1}$$

$$\mathbb{E} \left(\log \sum_i a_i \zeta_i \right) = \frac{1}{\delta} \log \mathbb{E}[\zeta^\delta]^{1/\delta}.$$

Suppose μ is as follows:

- a_j - P.D. parameter δ .
- $m_{u,j}$ IID copies of $S \in \mathcal{P}(\mathcal{P}(\mathcal{X}))$
 $\sim \text{Dir}(\dots)$

- $m_{u,j}$ $\perp \perp U$ copies of $S \in \mathcal{P}(\mathcal{P}(\mathcal{X}))$
 $\cong \mathcal{P}([0,1])$

$$\mu(x_u) = \sum_j a_j \prod_{u \in U} m_{u,j}(x_u)$$

Then

$$\begin{aligned} \Phi &\leq \mathbb{E} \log \left[\sum_j a_j \left(1 + \lambda \prod_{u,v} (m_{u,j}(0) + \delta m_{u,j}(1)) \right)^\delta \right] \\ &\quad - \frac{d}{2} \mathbb{E} \log \left[\left(\sum_j a_j \left(1 - (1-\delta) m_{u,j}(1) m_{u',j}(1) \right) \right)^\delta \right] \\ &= \delta^{-1} \log \mathbb{E} \left[\left(1 + \lambda \prod (1 - (1-\delta) m_u(1)) \right)^\delta \right] \\ &\quad - \delta^{-1} \frac{d}{2} \log \mathbb{E} \left[\left(1 - (1-\delta) m_u(1) m_{u'}(1) \right)^\delta \right] \end{aligned}$$

Recall: Sum over WP configurations r ,

$$Z = \sum w(r),$$

reweighted to those that exist,

$$Z_\delta = \sum w(r)^\delta.$$

When $\delta=0$ this is counting unweighted clusters.

$$SP_x[S](A) = \frac{1}{Z_\delta} \left[\mathbb{I} \left(\frac{\lambda \prod (1 - (1-\delta)r_i)}{1 + \lambda \prod (1 - (1-\delta)r_i)} \in A \right) \right]$$

$$SP_{\delta} [s] (A) = \frac{1}{\delta} \int I \left(\frac{\lambda \pi (1 - (1-\delta)r_i)}{1 + \lambda \pi (1 - (1-\delta)r_i)} \in A \right) \cdot \left(1 + \lambda \pi (1 - (1-\delta)r_i) \right)^{\delta} \prod_{i=1}^{d-1} s(dr_i)$$

Choose δ, s_{δ} according to this prediction.

• For MAX I.S. $\delta \rightarrow 0, \lambda \rightarrow \infty, \lambda^{\delta} \rightarrow \theta$

then $s_{\delta, \lambda, \theta} \xrightarrow{w} s_{\theta}$

our W.P. prediction with

• $s_{\theta}(1) = q, s_{\theta}(0) = 1 - q$

• $q = \frac{\theta (1 - q)^{d-1}}{1 + (\theta - 1) (1 - q)^{d-1}} .$