

Lecture 17: Spectral Independence from Zero-Freeness

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In this lecture, we return to (multivariate) zero-freeness of the partition function. We show that it implies spectral independence and hence, fast mixing of Glauber dynamics.

1 The Generating Polynomial and its Zero-Freeness

Let μ be a probability distribution over $2^{[n]}$ (which of course can be identified with $\{\pm 1\}^n$). Define its *generating polynomial* as

$$g_\mu(\mathbf{z}) \stackrel{\text{def}}{=} \sum_{S \subseteq [n]} \mu(S) \cdot \mathbf{z}^S, \quad (1)$$

where we use the shorthand $\mathbf{z}^S \stackrel{\text{def}}{=} \prod_{i \in S} z_i$. This is a *multiaffine* polynomial of degree (at most) n . Its logarithm is essentially the cumulant generating function \mathcal{L}_μ of μ we previously saw, except with a change of variables to obtain a polynomial. One of the central ideas in the study of the *geometry of polynomials* is the following.

Theorem 1.1. *It is fruitful to relate the analytic/algebraic properties of g_μ to the probabilistic/combinatorial properties of μ itself.*

For instance, a consequence of Barvinok's algorithm is that whenever g_μ admits a large zero-free region, then we have efficient algorithms for estimating g_μ .¹ In this lecture, we will be interested in deducing correlation bounds given zero-freeness of g_μ . The following beautiful theorem, which we will not prove, is an example of such a result.

Theorem 1.2 ([BBL09]). *Suppose g_μ is real stable, i.e. that $g_\mu \neq 0$ whenever $\text{Im } z_i > 0$ for all $i \in [n]$. Then μ is negatively correlated in the sense that*

$$\Pr_{S \sim \mu} [j \in S \mid i \in S] \leq \Pr_{S \sim \mu} [j \in S \mid i \notin S], \quad \forall i \neq j.$$

Real stability of the generating polynomial of μ is sometimes referred to as the *strongly Rayleigh property*. It turns out to be an extremely robust notion of negative correlation, one which has found many applications; see e.g. [Pem12] and references therein. We also previously saw that negatively correlated distributions on (homogeneous) set systems are 1-spectrally independent. The main result of this lecture is to establish a more direct connection between zero-freeness and spectral independence, one which does not require zero-freeness w.r.t. an entire half-plane.

Recall the following notion of multivariate zero-freeness we used previously.

Definition 1 (Stability). *Let $\Gamma_1, \dots, \Gamma_n \subseteq \mathbb{C}$ be subsets of the complex plane. We say a multivariate polynomial $p(z_1, \dots, z_n)$ is $\Gamma_1 \times \dots \times \Gamma_n$ -stable if $p(\mathbf{z}) \neq 0$ whenever $z_i \in \Gamma_i$ for all $i = 1, \dots, n$. If $\Gamma_1 = \dots = \Gamma_n = \Gamma$ for some $\Gamma \subseteq \mathbb{C}$, then we simply say p is Γ -stable.*

Theorem 1.3 ([CLV21]; building on [Ali+21]). *Suppose there exists a constant $\delta > 0$ such that g_μ is stable w.r.t. the open radius- δ disk $\mathbb{D}(1, \delta)$ around 1. If in addition the marginals of μ are bounded in the sense that there is a constant $0 < \mathcal{B} \leq 1/2$ such that $\Pr_{S \sim \mu} [i \in S], \Pr_{S \sim \mu} [i \notin S] \geq \mathcal{B}$ for all $i \in [n]$, then*

$$\sum_{j=1}^n |\Psi_\mu(i \rightarrow j)| \leq \frac{4}{\mathcal{B}(1 - \mathcal{B})\delta^2}, \quad \forall i \in [n].$$

In particular, μ is $O(1/\mathcal{B}\delta^2)$ -spectrally independent.

¹This is not entirely true, since Barvinok's algorithm also requires that that zero-free region contains a point at which computing g_μ is easy. For instance, we do not have FPTAS for estimating arbitrary real stable polynomials.

Remark 1. The point **1** for zero-freeness isn't special. One could look at stability w.r.t. an open radius- δ disk around any other point $\lambda \in \mathbb{R}_{\geq 0}^n$, in which case we'd get spectral independence for the tilted distribution $\mu_\lambda(S) \propto \mu(S) \cdot \lambda^S$.

Theorem 1.4 ([Ali+21]). *Suppose there exists a constant $\alpha > 0$ such that g_μ is stable w.r.t. the sector*

$$S_\alpha \stackrel{\text{def}}{=} \{re^{i\theta} : |\theta| < \alpha\pi/2, r > 0\}. \quad (2)$$

around the nonnegative real axis with aperture $\alpha\pi$. Then

$$\sum_{j=1}^n \left| \Pr_{S \sim \mu} [j \in S \mid i \in S] - \Pr_{S \sim \mu} [j \in S \mid i \notin S] \right| \leq \frac{2}{\alpha}, \quad \forall i \in [n],$$

and μ is $(\frac{2}{\alpha} - 1)$ -spectrally independent. Furthermore, the same inequality holds for all exponential tilts of μ , i.e. distributions of the form $\mu_\lambda(S) \propto \mu(S) \cdot \lambda^S$ for some $\lambda \in \mathbb{R}_{\geq 0}^n$.

Remark 2. In the special case $\alpha = 1$, S_α becomes the open right half-plane $\{z \in \mathbb{C} : \text{Re } z > 0\}$. Polynomials which are stable w.r.t. S_1 are often called *Hurwitz stable*.

The rough intuition behind these statements is the following: Since the correlations of μ are given by second-order derivatives of $\log g_\mu$ at **1**, these correlations are small if $\log g_\mu(\mathbf{1})$ is “smooth” in some sense. This is only the case if **1** is far away from the zeros of g_μ . For more results accommodating more general zero-free regions, see [Ali+21; CLV21].

1.1 Applications

Before we prove Theorems 1.3 and 1.4, let us mention a few applications.

Example 1 (Hardcore Model in Tree Uniqueness). We previously mentioned that Peters–Regts [PR19] established stability of the multivariate independence polynomial $Z_G(\lambda)$ of a graph of maximum degree Δ in a neighborhood of the interval $[0, \lambda_c(\Delta))$, where $\lambda_c(\Delta)$ is again the uniqueness threshold w.r.t. the infinite Δ -regular tree. In this regime, Barvinok’s algorithm furnishes an FPTAS. Combining this zero-freeness result with Theorem 1.3 yields an alternative proof of $O(1)$ -spectral independence of the hardcore Gibbs measure in the uniqueness regime, albeit with worse quantitative bounds.

Example 2 (Monomer-Dimer Model). Recall we previously showed that the univariate matching polynomial $\mathcal{M}_G(z) = \sum_{M \subseteq E \text{ matching}} z^{|M|}$ is real-rooted. This is the Heilmann–Lieb Theorem [HL72], and more in depth analysis reveals that for every nonnegative vector of edge weights $\lambda \in \mathbb{R}_{\geq 0}^E$, the multivariate (vertex) matching polynomial

$$\mathcal{M}_G(z) \stackrel{\text{def}}{=} \sum_{\substack{M \subseteq E \\ \text{matching}}} \prod_{e \in M} \lambda_e \prod_{v \text{ unmatched}} z_v,$$

is *Hurwitz stable*; see e.g. [BB09]. This in particular implies that for any graph $G = (V, E)$, the monomer-dimer Gibbs distribution satisfies the correlation bounds

$$\sum_{v \in V} \left| \Pr_M[v \text{ matched} \mid u \text{ matched}] - \Pr_M[v \text{ matched} \mid u \text{ unmatched}] \right| \leq 2.$$

Note that no assumptions on the degree were made.

There are also additional applications to determinantal point processes, even subgraphs, edge covers, antiferromagnetic spin systems on line graphs, etc. [Ali+21; CLV21].

2 A Little Complex Analysis

To formalize the above intuition, we will leverage the following standard fact from complex analysis, which captures the “rigidity” of smooth complex functions.

Lemma 2.1 (Schwarz–Pick). *Let $f : \mathbb{D}(0, 1) \rightarrow \mathbb{D}(0, 1)$ is a univariate holomorphic function. Then $|f'(0)| \leq 1 - |f(0)|^2 \leq 1$.*

In light of the Schwarz–Pick lemma, our strategy will be to construct such a univariate holomorphic function f such that f maps $\mathbb{D}(0, 1)$ into itself, and $|f'(0)| \approx \sum_{j=1}^n |\Psi_\mu(i \rightarrow j)|$, perhaps up to constants depending on \mathcal{B}, δ .

Proof of Theorem 1.3. Fix an arbitrary $i \in [n]$, and define

$$F_i(\mathbf{z}) \stackrel{\text{def}}{=} \frac{\partial_{z_i} \log g_\mu(\mathbf{z})}{\Pr_{S \sim \mu}[i \in S] \cdot \Pr_{S \sim \mu}[i \notin S]}.$$

Note that $\partial_{z_j} F_i(\mathbf{1}) = \Psi_\mu(i \rightarrow j)$ for all $j \in [n]$. Our goal is to construct appropriate maps $\psi : \mathbb{C} \rightarrow \mathbb{C}$ and $\varphi : \mathbb{C} \rightarrow \mathbb{C}^n$ such that the composition $f(z) = \psi(F_i(\varphi_1(z), \dots, \varphi_n(z)))$ is holomorphic and satisfies $f(\mathbb{D}(0, 1)) \subseteq \mathbb{D}(0, 1)$. If we have such ψ, φ , then by applying the Schwarz–Pick Lemma,

$$\begin{aligned} 1 &\geq |f'(0)| && \text{(Lemma 2.1)} \\ &= |\psi'(F_i(\varphi(0)))| \cdot \langle \nabla F_i(\varphi(0)), \varphi'(0) \rangle && \text{(Chain Rule)} \\ &= |\psi'(F_i(\varphi(0)))| \cdot \sum_{j=1}^n \Psi_\mu(i \rightarrow j) \cdot \varphi'_j(0). \end{aligned}$$

A natural and simple choice is to take ψ, φ to be *affine* functions.

- To ensure that f is holomorphic, we use φ to map $\mathbb{D}(0, 1)$ into the region of stability of g_μ , which we assumed is $\mathbb{D}(1, \delta)$. For convenience, let us allow an epsilon of room. Let

$$\varphi_j(z) \stackrel{\text{def}}{=} 1 + \frac{\delta}{2} s_j z,$$

where $s_j = \text{sign}(\Psi_\mu(i \rightarrow j))$. This ensures that $\varphi_j(\mathbb{D}(0, 1)) \subseteq \mathbb{D}(1, \delta/2)$ for all $j \in [n]$,

$$\sum_{j=1}^n \Psi_\mu(i \rightarrow j) \cdot \varphi'_j(0) = \frac{\delta}{2} \cdot \sum_{j=1}^n |\Psi_\mu(i \rightarrow j)|.$$

In particular, this choice for φ alone already implies

$$\sum_{j=1}^n |\Psi_\mu(i \rightarrow j)| \leq \frac{2}{\delta \cdot |\psi'(F_i(\mathbf{1}))|}. \quad (3)$$

- Now, we design ψ . In order to apply the Schwarz–Pick Lemma, we need ψ to map the image of $\mathbb{D}(0, 1)$ under $F_i \circ \varphi$ back into $\mathbb{D}(0, 1)$. This is the tricky part, since we must understand the image of F_i . This is also where it is convenient to have an epsilon of room from the definition of φ , since we only need to consider $F_i(\mathbb{D}(1, \delta/2))$ instead of $F_i(\mathbb{D}(1, \delta))$.

Claim 2.2. *The image of $\mathbb{D}(1, \delta/2)$ under F_i is contained in $\mathbb{D}\left(0, \frac{2}{\mathcal{B}(1-\mathcal{B})\delta}\right)$.*

Once we have this, an obvious choice for ψ is to scale everything down by a factor of $\frac{\mathcal{B}(1-\mathcal{B})\delta}{2}$. In particular, we let $\psi(z) = \frac{\mathcal{B}(1-\mathcal{B})\delta}{2} z$. Plugging this into Eq. (3) immediately implies the theorem. All that remains is to justify Claim 2.2. □

Proof of Claim 2.2. Since we assumed the marginal bound $\Pr_{S \sim \mu}[i \in S] \geq \mathcal{B}$, the claim is equivalent to showing that the image of $\mathbb{D}(1, \delta/2)$ under $\mathbf{z} \mapsto \partial_i \log g_\mu(\mathbf{z})$ is contained in $\mathbb{D}(0, 2/\delta)$.

We go by contradiction, making crucial use linearity of g_μ in each of its variables. Fix $z_1, \dots, z_n \in \mathbb{D}(1, \delta/2)$, and write $y = \partial_i \log g_\mu(\mathbf{z})$. We wish to show that $|y| < 2/\delta$. Since

$$y = \partial_i \log g_\mu(\mathbf{z}) = \frac{\partial_i g_\mu(\mathbf{z})}{g_\mu(\mathbf{z})},$$

rearranging yields

$$g_\mu(\mathbf{z}) - \frac{1}{y} \cdot \partial_i g_\mu(\mathbf{z}) = 0. \quad (4)$$

Suppose for contradiction that $|y| \geq \frac{2}{\delta}$. Then $\left| -\frac{1}{y} \right| \leq \frac{\delta}{2}$. We use this and [Eq. \(4\)](#) to construct a new vector $\mathbf{z}' \in \mathbb{D}(1, \delta)^n$ such that $g_\mu(\mathbf{z}') = 0$, contradicting $\mathbb{D}(1, \delta)$ -stability of g_μ .

Define \mathbf{z}' by $z'_j = z_j$ for all $j \neq i$, and $z'_i = z_i - \frac{1}{y}$. Since $|z_i - 1| < \frac{\delta}{2}$ and $\left| -\frac{1}{y} \right| \leq \frac{\delta}{2}$, we have $\mathbf{z}' \in \mathbb{D}(1, \delta)^n$. Furthermore, since g_μ is linear in each of its variables,

$$\begin{aligned} g_\mu(\mathbf{z}') &= \underbrace{(g_\mu(\mathbf{z}) - z_i \cdot \partial_i g_\mu(\mathbf{z}))}_{\text{Monomials without } i} + \underbrace{\left(z_i - \frac{1}{y} \right) \cdot \partial_i g_\mu(\mathbf{z})}_{\text{Monomials with } i} \\ &= g_\mu(\mathbf{z}) - \frac{1}{y} \cdot \partial_i g_\mu(\mathbf{z}) \\ &= 0 \end{aligned} \quad (\text{By [Eq. \(4\)](#)})$$

□

3 Better Maps for Sectors

If we impose more structure on our zero-free regions, then we can construct much better ψ, φ and prove better bounds. We just need to understand where how the map $y \mapsto -\frac{1}{y}$ changes our zero-free region, and how to map between these regions and the unit disk $\mathbb{D}(0, 1)$.

Proof of [Theorem 1.4](#). Since μ and μ_λ have the same generating polynomials up to rescaling the variables by nonnegative coefficients, sector stability also holds for g_{μ_λ} . Hence, without loss of generality, we just prove spectral independence for μ itself.

Fix an arbitrary $i \in [n]$, and define²

$$F_i(\mathbf{z}) = \log \left(\frac{\partial_i g_\mu(\mathbf{z})}{(1 - z_i \partial_i) g_\mu(\mathbf{z})} \right).$$

A direct calculation reveals that $\partial_{z_j} F_i(\mathbf{1}) = \Pr_{S \sim \mu}[j \in S \mid i \in S] - \Pr_{S \sim \mu}[j \in S \mid i \notin S]$ for all $j \in [n]$. We construct appropriate maps $\psi : \mathbb{C} \rightarrow \mathbb{C}$ and $\varphi : \mathbb{C} \rightarrow \mathbb{C}^n$ such that the composition $f(z) = \psi(F_i(\varphi_1(z), \dots, \varphi_n(z)))$ satisfies the assumptions of the Schwarz–Pick Lemma, since then we’d have

$$1 \geq |f'(0)| = |\psi'(F_i(\varphi(0)))| \cdot \sum_{j \neq i} \Psi_\mu(i \rightarrow j) \cdot \varphi'_j(0). \quad (5)$$

- Since g_μ is stable w.r.t. the sector S_α , we use Möbius transformations and exponential maps instead of affine functions. More specifically, take

$$\begin{aligned} \varphi_j(z) &\stackrel{\text{def}}{=} g(s_j z)^\alpha = \left(\frac{1 + s_j z}{1 - s_j z} \right)^\alpha \\ \text{where } g(x) &= \frac{1 + x}{1 - x} \quad \text{and} \quad s_j = \text{sign}(\Psi_\mu(i \rightarrow j)), \quad \forall j \in [n]. \end{aligned} \quad (6)$$

The point is that $\varphi_j(\mathbb{D}(0, 1)) \subseteq S_\alpha$ since the inner Möbius function g maps $\mathbb{D}(0, 1)$ to the right half-plane S_1 , and then taking the α th power scales down the angle. A quick calculation reveals that $\varphi'_j(z) = 2s_j \alpha \cdot \left(\frac{1 + s_j z}{1 - s_j z} \right)^{\alpha-1} \cdot \frac{1}{(1 - s_j z)^2}$ and so plugging this into [Eq. \(5\)](#) and using $\varphi(0) = \mathbf{1}$ gives

$$\sum_{j \neq i} |\Psi_\mu(i \rightarrow j)| \leq \frac{1}{2\alpha} \cdot \frac{1 - \psi'(F_i(\mathbf{1}))^2}{|\psi'(F_i(\mathbf{1}))|}. \quad (7)$$

- Now let us argue about the image of F_i , which will then tell us how to construct ψ .

²If the numerator inside the logarithm were multiplied by z_i , and if $\mathbf{z} \in \mathbb{R}_{\geq 0}^n$, then we’d exactly have the marginal ratio of i under the tilted measure $\mu_{\mathbf{z}}$.

Claim 3.1. *For every $z_1, \dots, z_n \in S_\alpha$, we have that*

$$\frac{\partial_i g_\mu(z)}{(1 - z_i \partial_i) g_\mu(z)} \notin -S_\alpha.$$

In particular, the image of S_α under F_i is contained within the strip

$$\left\{ z \in \mathbb{C} : |\operatorname{Im} z| < \left(1 - \frac{\alpha}{2}\right) \pi \right\}. \quad (8)$$

Before we prove this claim, let us finish the proof by constructing ψ . Let

$$\psi(z) = g^{-1} \left(\exp \left(\frac{1/2}{1 - \alpha/2} \cdot z \right) \right),$$

where $g^{-1}(z) = \frac{z-1}{z+1}$ is the inverse of the Möbius transformation g we used in the definition of φ above. The point is that the inner exponential maps the strip in [Claim 3.1](#) to the right half-plane S_1 , and then g^{-1} maps this right half-plane to $\mathbb{D}(0, 1)$. Another quick calculation reveals that

$$\psi'(z) = \frac{2 \exp \left(\frac{1/2}{1 - \alpha/2} \cdot z \right)}{\left(1 + \exp \left(\frac{1/2}{1 - \alpha/2} \cdot z \right) \right)^2} \cdot \frac{1/2}{1 - \alpha/2} = \frac{1/2}{1 - \alpha/2} \cdot \frac{1}{2} (1 - \psi(z)^2).$$

Combined with [Eq. \(7\)](#) yields

$$\sum_{j \neq i} |\Psi_\mu(i \rightarrow j)| \leq \frac{2}{\alpha} - 1.$$

Adding back $\Psi_\mu(i \rightarrow i) = 1$ to both sides concludes the proof. □

Proof of [Claim 3.1](#). Since the image of $-S_\alpha$ under the exponential map $z \mapsto \exp(z)$ is the strip in [Eq. \(8\)](#), the first claim indeed implies the second. Let $y = \frac{\partial_i g_\mu(z)}{(1 - z_i \partial_i) g_\mu(z)}$ and suppose for contradiction that $y \in S_{-\alpha}$. Then $-\frac{1}{y} \in S_\alpha$, whence

$$\begin{aligned} 0 &= (1 - z_i \partial_i) g_\mu(z) - \frac{1}{y} \cdot \partial_i g_\mu(z) && \text{(Rearranging)} \\ &= g_\mu \left(-\frac{1}{y}, z_{-i} \right). && (g_\mu \text{ is multiaffine}) \end{aligned}$$

This contradicts S_α -stability of g_μ . □

References

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