



Biased Random walk on partitions and hyperoctahedral Hall-Littlewood Polynomials.

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Content



Introduction

Solution via orthogonal polynomials

Discrete Laplacian

Bernstein–Szegő Polynomials

Multiple variables

Laplacian on $\Lambda^{(N)}$

Hall–Littlewood Polynomial

Content



Introduction

Solution via orthogonal polynomials

Discrete Laplacian

Bernstein–Szegő Polynomials

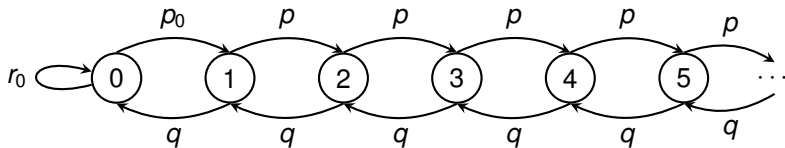
Multiple variables

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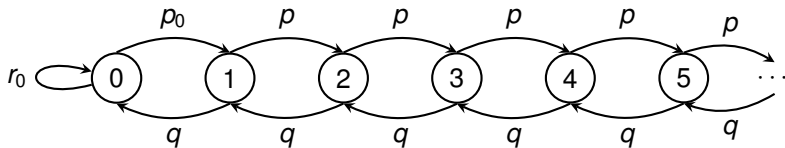
Biased random walk with a reflecting barrier

The motion of a particle in a discrete space can be stochastically modeled as a Markov chain. We also consider a reflecting boundary at node 0.



Biased random walk with a reflecting barrier

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With $p + q = 1$ and $p_0 + r_0 = 1$. This class of problems has been widely studied from a different perspective than the one we will consider today, see [1].

Transition matrix

Let $p_0 = p$ and $r_0 = q$. Then we define the Markov chain $\{X_n\}_{n \in \mathbb{N}_0}$ through the transition probability matrix

$$\mathbf{P} = \begin{bmatrix} q & p & 0 & 0 & \cdots \\ q & 0 & p & 0 & \\ 0 & q & 0 & p & \\ \vdots & & \ddots & \ddots & \ddots \end{bmatrix}, \quad \text{where } q = \frac{b}{b + b^{-1}} \text{ and } p = \frac{b^{-1}}{b + b^{-1}}. \quad (1)$$

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Here $\mathbf{P}_{i,j}$ represents the probability of reaching state $j - 1$ from state $i - 1$. Usually, the objective is to find the transition probabilities after n steps, which means computing $\mathbf{P}_{i,j}^n$.

Interpretation as a operator on \mathbb{N}_0

Let $\mu \in \ell^2(\mathbb{N}_0)$, one can define $\mathbf{P} : \ell^2(\mathbb{N}_0) \rightarrow \ell^2(\mathbb{N}_0)$ as:

$$\mathbf{P}(\mu) = (\mathbf{P}\mu)(n) = \begin{cases} \frac{b}{b+b^{-1}} \varphi(n-1) + \frac{b^{-1}}{b+b^{-1}} \varphi(n+1), & n > 0 \\ \frac{b}{b+b^{-1}} \varphi(0) + \frac{b^{-1}}{b+b^{-1}} \varphi(1), & n = 0 \end{cases} \quad (2)$$

with $b \in (0, \infty)$. Diagonalizing \mathbf{P} opens the door to answering different probabilistic questions about the Markov chain $\{X_n\}_{n \in \mathbb{N}_0}$.

Probabilistic questions



Once the Markov chain is defined, several probabilistic quantities become relevant:

- ▶ Transition probabilities after n steps:

$$\mathbb{P}(X_n = j \mid X_0 = i) = \mathbf{P}_{ij}^n.$$

- ▶ Spectral data of the transition operator: eigenvalues and eigenvectors of \mathbf{P} .
- ▶ Long-time behavior of the chain.
- ▶ Stationary distributions and recurrence properties.
- ▶ Mixing behavior and convergence rates.

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Orthogonal polynomials provide an efficient way to study these quantities through spectral decomposition.

Content



Introduction

Solution via orthogonal polynomials

Discrete Laplacian

Bernstein–Szegő Polynomials

Multiple variables

Laplacian on $\Lambda^{(N)}$

Hall–Littlewood Polynomial

Content



Introduction

Solution via orthogonal polynomials

Discrete Laplacian

Bernstein–Szegő Polynomials

Multiple variables

Laplacian on $\Lambda^{(N)}$

Hall–Littlewood Polynomial

Discrete Laplacian on \mathbb{N}_0



Let $\varphi \in \ell^2(\mathbb{N}_0)$,

$$L(\varphi) = (L\varphi)(n) = \begin{cases} \varphi(n-1) + \varphi(n+1), & n > 0 \\ b\varphi(0) + (1-a)\varphi(1), & n = 0 \end{cases},$$

with $a, b \in \mathbb{R}$ and $a < 1$.

Discrete Laplacian on \mathbb{N}_0



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with $a, b \in \mathbb{R}$ and $a < 1$. The spectral equation

$$L\varphi = E\varphi, \tag{3}$$

has formal solutions for $E = 2 \cos(\xi)$, for each $\xi \in [0, \pi]$. Let ψ_ξ be of the following form:

$$\psi_\xi := \psi_\xi(n) = U_n(\cos(\xi)) - bU_{n-1}(\cos(\xi)) + aU_{n-2}(\cos(\xi)),$$

where $U_n(\cos(\xi)) = \frac{\sin((n+1)\xi)}{\sin(\xi)}$. Then ψ_ξ is a solution of (3).

Content



Introduction

Solution via orthogonal polynomials

Discrete Laplacian

Bernstein–Szegő Polynomials

Multiple variables

Laplacian on $\Lambda^{(N)}$

Hall–Littlewood Polynomial

Bernstein–Szegő polynomials

The functions $\psi_\xi(n)$ can be written as bi-parametric Bernstein–Szegő polynomials,

$$\psi_\xi(n) = B_n^{(\alpha, \beta)}(z) = c(z; \alpha, \beta)z^n + c(z^{-1}; \alpha, \beta)z^{-n},$$

where

$$z = e^{i\xi}, \quad c(z; \alpha, \beta) := \frac{(1 - \alpha z^{-1})(1 - \beta z^{-1})}{1 - z^{-2}},$$

with the relations $a = \alpha\beta$ and $b = \alpha + \beta$.

Bernstein–Szegő polynomials



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with the relations $a = \alpha\beta$ and $b = \alpha + \beta$. From now on we only consider the parameter region relevant to our work ($a = 0$). Thus, we set $\beta = 0$, which means

$$B_n^{(\alpha, \beta)} = B_n^{(\alpha)}, \quad c(z; \alpha, \beta) = c(z; \alpha) = \frac{1 - \alpha z^{-1}}{1 - z^{-2}}. \quad (4)$$

Orthogonality relation

Proposition

Let $\alpha \in \mathbb{C}$. Then the following orthogonality relation holds for all $l, k \in \mathbb{N}_0$:

$$\begin{aligned} & \frac{1}{2\pi} \int_0^\pi B_l^{(\alpha)}(e^{i\xi}) B_k^{(\alpha)}(e^{i\xi}) \Delta(\xi; \alpha) d\xi + B_l^{(\alpha)}(\alpha) B_k^{(\alpha)}(\alpha) \Delta_\alpha \\ &= \begin{cases} 1 & \text{if } l = k, \\ 0 & \text{if } l \neq k, \end{cases} \end{aligned}$$

with

$$\Delta(\xi; \alpha) = \frac{1}{c(e^{i\xi}; \alpha) c(e^{-i\xi}; \alpha)}, \quad \text{and} \quad \Delta_\alpha = \begin{cases} \frac{\alpha - \alpha^{-1}}{\alpha} & \text{if } |\alpha| > 1, \\ 0 & \text{if } |\alpha| \leq 1. \end{cases}$$

This formula is a special case of the result in [3].

Spectrum of \mathbf{P}

From here we can deduce the explicit eigenfunctions of \mathbf{P} :

$$Q(\xi) = \begin{bmatrix} Q_0(\xi) \\ \vdots \\ Q_j(\xi) \\ \vdots \end{bmatrix} \quad \text{with} \quad Q_j(\xi) = b^j B_j^{(\alpha)}(e^{i\xi}),$$

and the associated eigenvalues are given by:

$$E(\xi) = \begin{cases} \frac{2 \cos(\xi)}{b+b^{-1}} & \xi \in [0, \pi], \\ 1 & \xi = i \ln(b). \end{cases}$$

Content



Introduction

Solution via orthogonal polynomials

Discrete Laplacian

Bernstein–Szegő Polynomials

Multiple variables

Laplacian on $\Lambda^{(N)}$

Hall–Littlewood Polynomial

Content



Introduction

Solution via orthogonal polynomials

Discrete Laplacian

Bernstein–Szegő Polynomials

Multiple variables

Laplacian on $\Lambda^{(N)}$

Hall–Littlewood Polynomial

Generalization to N variables



It is possible to define the Markov chain $\{X_n^{(N)}\}_{n \in \mathbb{N}_0}$, where $X_n^{(N)} \in \Lambda^{(N)} := \left\{ \lambda = (\lambda_1, \dots, \lambda_N) \in (\mathbb{N}_0)^N \mid \lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_N \geq 0 \right\}$. We need to define:

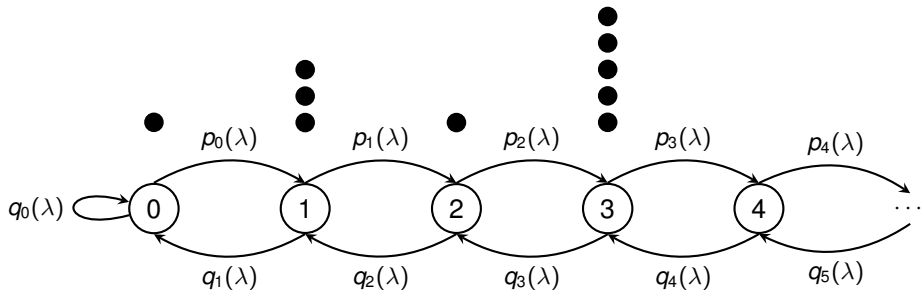
▶ $m_l(\lambda) := |\{j \mid \lambda_j = l; \quad j = 1, \dots, n\}|$.

▶ $[n]_t := \frac{1-t^n}{1-t} = \begin{cases} 0, & \text{if } n = 0, \\ 1 + t + t^2 + \dots + t^{n-1}, & \text{if } n > 0. \end{cases}$

▶ $\vartheta_l(\lambda) := \max_{j \in \mathbb{N}_0} \{j \mid \lambda_j = l\}$.

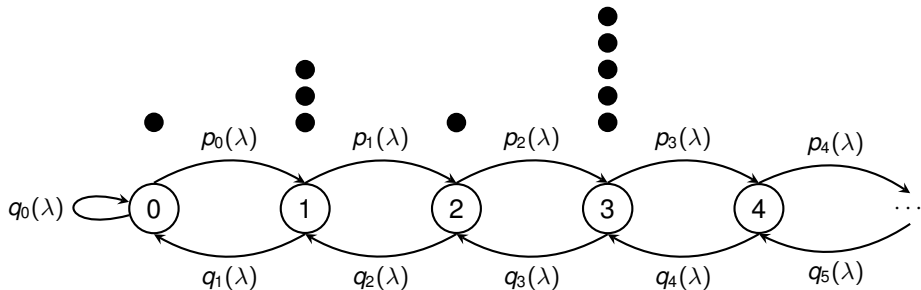
Generalization to N variables

Intuitively, the Markov chain $\{X_n^{(N)}\}_{n \in \mathbb{N}_0}$ follows the next diagram at each step:



Generalization to N variables

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The probabilities are defined by

$$p_l(\lambda) = \frac{1}{\Sigma} \left(b^{-1} t^{-(n-\vartheta_l(\lambda))} [m_l(\lambda)]_{t^{-1}} \right), \quad q_l(\lambda) = \frac{1}{\Sigma} \left(b t^{(n-\vartheta_l(\lambda))} [m_l(\lambda)]_t \right),$$

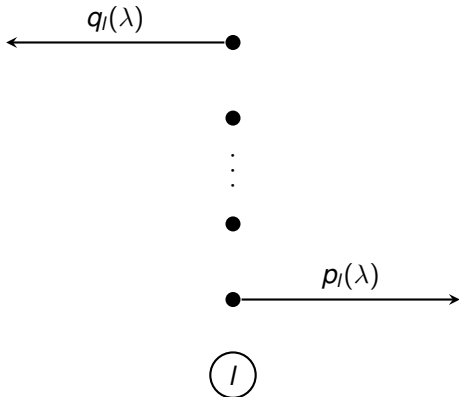
with normalization constant

$$\Sigma = b[N]_t + b^{-1}[N]_{t^{-1}}.$$

The model can be interpreted as a system of N ordered particles on \mathbb{N}_0 .

Generalization to N variables

If we focus on a single node, $l = \lambda_{\vartheta_l(\lambda)} = \lambda_{\vartheta_l(\lambda)-1} = \cdots = \lambda_{\vartheta_l(\lambda)-m_l(\lambda)+1}$,



Laplacian on $\Lambda^{(N)}$

Let $\varphi \in \ell^2(\Lambda^{(N)})$. We define $\mathbf{P}^{(N)}$ as

$$\mathbf{P}^{(N)}(\varphi) = \left(\begin{aligned} & \sum_{\substack{1 \leq j \leq N \\ \lambda + \mathbf{e}_j \in \Lambda^{(N)}}} b^{-1} t^{-(n-j)} [m_{\lambda_j}(\lambda)]_t \varphi_{\lambda + \mathbf{e}_j} \\ & + \sum_{\substack{1 \leq j \leq N \\ \lambda - \mathbf{e}_j \in \Lambda^{(N)}}} b t^{(n-j)} [m_{\lambda_j}(\lambda)]_t \varphi_{\lambda - \mathbf{e}_j} + \delta_{\lambda_N} b t^{(n-j)} [m_{\lambda_N}(\lambda)]_t \varphi_{\lambda} \end{aligned} \right) \frac{1}{\Sigma}.$$

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which is a generalization of (2), and coincides, up to normalization and parameter identification, with the n -particle Hamiltonian of the semi-infinite q -boson system studied in [2].

Content



Introduction

Solution via orthogonal polynomials

Discrete Laplacian

Bernstein–Szegő Polynomials

Multiple variables

Laplacian on $\Lambda^{(N)}$

Hall–Littlewood Polynomial

Hall–Littlewood polynomial with hyperoctahedral symmetry



Let $\mathbf{z} = (z_1, \dots, z_N)$, $t \in [-1, 1]$, $\alpha \in \mathbb{C}$ and $\lambda \in \Lambda^{(N)}$. We define a Hall–Littlewood type polynomial with hyperoctahedral symmetry by

$$R_\lambda(\mathbf{z}; \alpha, t) := \sum_{\substack{\sigma \in \mathcal{S}_N \\ \epsilon \in \{-1, 1\}^N}} C(\mathbf{z}_\sigma^\epsilon; \alpha, t) \prod_{i=1}^N z_{\sigma(i)}^{\epsilon_i \lambda_i}, \quad (5)$$

where

$$C(\mathbf{z}; \alpha, t) = \prod_{i=1}^N c(z_i; \alpha) \prod_{1 \leq i < j \leq N} \frac{z_i - tz_j}{z_i - z_j} \frac{z_i z_j - t}{z_i z_j - 1}.$$

Hall–Littlewood polynomial with hyperoctahedral symmetry



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The polynomial $B_n^{(\alpha)}(z)$ arises as the special case $N = 1$. For the specialization $t = 0$, one obtains a family of polynomials that generalizes the Schur polynomials. These polynomials arise in the diagonalization of another operator and satisfy a natural orthogonality relation (see [4]).

Orthogonality relation



The family $\{R_\lambda\}_{\lambda \in \Lambda^{(N)}}$, with the change of variable $(z_1, \dots, z_N) = (e^{i\xi_1}, \dots, e^{i\xi_N}) = e^{i\xi}$, forms an orthogonal basis of $L^2([0, \pi]^N, \Delta^{(N)})$.

Orthogonality relation

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Proposition

Let $|\alpha| < 1$, $t \in [-1, 1]$. Then the following orthogonality relation holds for all $\lambda, \mu \in \Lambda^{(N)}$:

$$\int_{[0, \pi]^N} R_\lambda(e^{i\xi}; \alpha, t) R_\mu(e^{i\xi}; \alpha, t) \Delta^{(N)} d\xi_1 \dots d\xi_N = \begin{cases} N(\lambda) & \text{if } \lambda = \mu, \\ 0 & \text{if } \lambda \neq \mu. \end{cases}$$

where

$$\Delta^{(N)} := \frac{1}{(2\pi)^N |C(e^{i\xi}; \alpha, t)|^2}.$$

Spectral data



From here we have the spectral decomposition of $\mathbf{P}^{(N)}$:

$$\mathbf{P}^{(N)}Q^{(N)}(\mathbf{z}) = E^{(N)}(\mathbf{z})Q^{(N)}(\mathbf{z}),$$

where

$$Q^{(N)}(\mathbf{z}) = Q_{\lambda}^{(N)}(\mathbf{z}) = \frac{R_{\lambda}(z_1, z_2, \dots, z_N)}{R_{\lambda}(\alpha t^0, \alpha t, \dots, \alpha t^{N-1})},$$

and

$$E^{(N)}(\mathbf{z}) = \sum_{j=1}^N \frac{z_j + z_j^{-1}}{\Sigma}.$$

Transition probabilities



Using the spectral decomposition of the operator $\mathbf{P}^{(N)}$, the transition probabilities of the Markov chain can be expressed in terms of the eigenfunctions R_λ .

$$\left(\mathbf{P}^{(N)}\right)_{\lambda,\mu}^m = \int_{[0,\pi]^N} \left(E^{(N)}(e^{i\xi})\right)^m Q_\lambda(e^{i\xi}) Q_\mu(e^{i\xi}) \Delta^{(N)} d\xi_1, \dots, \xi_N.$$

This representation allows us to study

- ▶ the long-time behavior of the Markov chain,
- ▶ convergence to equilibrium,
- ▶ mixing behavior and convergence rates,
- ▶ study the problem by generalizing the parameter regime.

Bibliography



- [1] M. DOMÍNGUEZ DE LA IGLESIA, *Orthogonal Polynomials in the Spectral Analysis of Markov Processes: Birth-Death Models and Diffusion*, Encyclopedia of Mathematics and Its Applications, Cambridge University Press, 2021.
- [2] J. F. VAN DIEJEN AND E. EMSIZ, *The semi-infinite q -boson system with boundary interaction*, Letters in Mathematical Physics, 104 (2014), pp. 103–113.
- [3] J. F. VAN DIEJEN, A. SOLEDISPA TIBÁN, AND A. VIDAL, *The spectrum of the finite open xx quantum spin chain with transverse magnetic boundary fields via orthogonal polynomials*, Annales Henri Poincaré, (2025).
- [4] A. VIDAL, *El espectro del laplaciano discreto en \mathbb{Z}_+ , mediante la ortogonalidad de polinomios de bernstein-szegő, y funciones de schur generalizadas*, master's thesis, Universidad de Talca, Talca, Chile, 2023.



Thank you for your attention.