

# The cutoff phenomenon for exclusion processes on graphs with open boundaries

Joe P. Chen

Department of Mathematics  
Colgate University  
Hamilton, NY, USA

The City College of New York  
Mathematics Colloquium  
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COLGATE UNIVERSITY

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Motivations  
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Convergence  
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Cutoff  
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Exclusion  
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Exclusion cutoffs  
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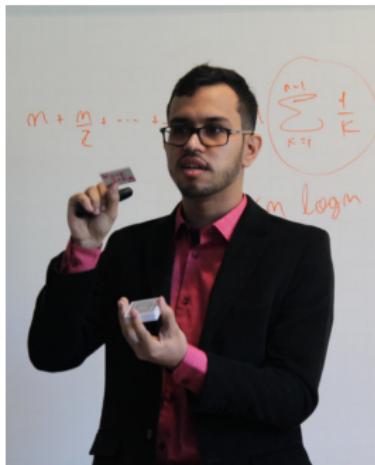
Proof ideas  
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Coda  
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This is based on joint work with



**Milton Jara**  
(IMPA)



**Rodrigo Marinho**  
(Técnico Lisboa)

Stay tuned for our preprint coming soon on arXiv!

# Outline

## Motivations

Convergence of Markov chains

The cutoff phenomenon

The exclusion process on graphs with boundaries

Cutoffs for exclusion processes

Proof ideas

Thank you for listening!

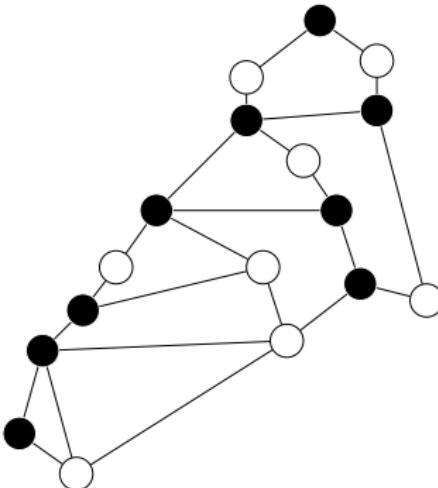
## Mixing a deck of cards



- Deck with  $n$  cards  $\Rightarrow n!$  permutations
- Deck with 52 cards  $\Rightarrow \sim 10^{68}$  permutations
- A deck will be considered **mixed** whenever all permutations are equally probable.

**Question:** How long does it take to mix a deck of 52 cards, using your preferred shuffle method?

## Mixing road traffic



- ● = Position occupied by a vehicle; ○ = Position vacant.
- A vehicle is allowed to move to an adjacent position provided that it is vacant.

**Questions:** How long does it take to mix the vehicles on this road network, according to the random adjacent moves?

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## Markov chains

- Fix a finite state space  $\Omega$ . To each  $x, y \in \Omega$ ,  $x \neq y$ , we assign a transition rate  $Q(x, y) \geq 0$ , and define  $Q(x, x) = -\sum_{y \neq x} Q(x, y) =: -\lambda_x$ .
  - The continuous-time Markov chain (MC) with rate matrix  $Q$  — “hold-jump” construction
- 
- $Q$  is the infinitesimal generator (“Laplacian”) for the Markov chain. A negative semidefinite operator.
  - If  $P_t(x, y)$  denotes the probability of moving from  $x$  to  $y$  over time  $t \geq 0$ , then it satisfies the backward/forward Kolmogorov’s equation  $\partial_t P_t(x, y) = (QP_t)(x, y) = (P_t Q)(x, y)$ .  
 $\implies P_t(x, y) = e^{tQ} P_0(x, y)$

## Convergence of Markov chains

- **Ergodic theorem for MCs.** Under irreducibility, there exists a unique prob. measure  $\pi$  (stationary distribution) s.t.  $\lim_{t \rightarrow +\infty} P_t(x, y) = \pi(y)$  for all  $x, y \in \Omega$ .
- **Can we state this convergence more quantitatively?**

Assume that  $Q$  is self-adjoint w.r.t.  $\pi$  (that is, the MC is reversible). By linear algebra,  $-Q =: L$  has all  $\mathbb{R}$ -valued eigenvalues  $\{\lambda_i\}_{i=0}^{|\Omega|-1}$ , listed in increasing order; and corresponding  $L^2(\pi)$ -orthonormal eigenfunctions  $\{\psi_i\}_{i=0}^{|\Omega|-1}$ .

Spectral decomposition      
$$(e^{tQ} f)(x) = \sum_{i=0}^{|\Omega|-1} e^{-t\lambda_i} \langle f, \psi_i \rangle_{L^2(\pi)} \psi_i(x).$$

- $\lambda_0 = 0$  ALWAYS. ( $Q$  is a singular matrix.)
- Rate of convergence is governed by  $\lambda_1$  (the spectral gap), which is positive by irreducibility and Perron-Frobenius.

## Convergence of Markov chains

- **Ergodic theorem for MCs.** Under irreducibility, there exists a unique prob. measure  $\pi$  (stationary distribution) s.t.  $\lim_{t \rightarrow +\infty} P_t(x, y) = \pi(y)$  for all  $x, y \in \Omega$ .
- **Can we state this convergence more quantitatively?**  
Use the **total variation distance** between two probability measures  $\mu$  and  $\nu$ :

$$\|\mu - \nu\|_{\text{TV}} := \sup_{S \subset \Omega} |\mu(S) - \nu(S)| = \frac{1}{2} \sum_{x \in \Omega} |\mu(x) - \nu(x)| \in [0, 1].$$

### TV distance to stationarity

$$d(t) := \sup_{x \in \Omega} \|P_t(x, \cdot) - \pi\|_{\text{TV}} = \sup_{x \in \Omega} \|(e^{tQ} P_0)(x, \cdot) - \pi\|_{\text{TV}}$$

We shall be interested in how  $d(t)$  drops from 1 to 0 as  $t$  increases.

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The exclusion process on graphs with boundaries

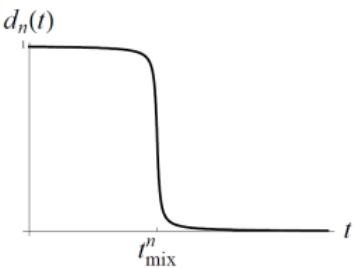
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## Cutoff := sharp transition in convergence to stationarity

- Let  $\{\{\eta_t^N : t \geq 0\}\}_{N \in \mathbb{N}}$  be a family of irreducible MCs, each of which has state space  $\Omega_N$ , with  $|\Omega_N| \uparrow \infty$ .
- $\pi^N$ : Stationary distribution of the  $N$ th chain.
- $d_N(t) := \sup_{\eta \in \Omega_N} \|\mathbb{P}_\eta(\eta_t^N \in \cdot) - \pi^N\|_{\text{TV}}$ : distance to stationarity
- A typical graph of  $t \mapsto d_N(t)$  looks like this:

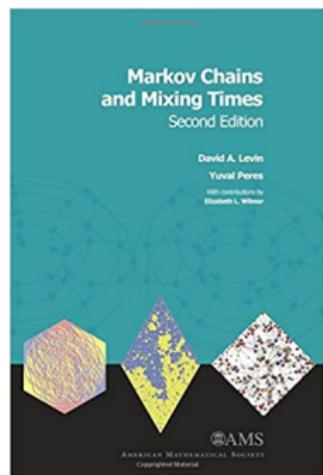
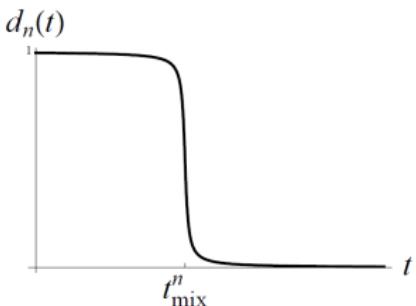


- $t_{\text{mix}}^N(\varepsilon) := \inf\{t \geq 0 : d_N(t) \leq \varepsilon\}$ .  **$\varepsilon$ -mixing time**.  
Some books fix  $\varepsilon = \frac{1}{4}$ . For us the full range  $\varepsilon \in [0, 1]$  is needed.

Cutoff := sharp transition in convergence to stationarity

**Def.**  $\{\{\eta_t^N : t \geq 0\}\}_{N \in \mathbb{N}}$  is said to exhibit **total variation cutoff** at times  $\{t_N\}_{N \in \mathbb{N}}$  if:

- For every  $\varepsilon \in (0, 1)$ ,  $\lim_{N \rightarrow \infty} \frac{t_{\text{mix}}^N(\varepsilon)}{t_N} = 1$ .
- Equivalently,  $\lim_{N \rightarrow \infty} d_N(\kappa t_N) = \begin{cases} 1, & \text{if } \kappa < 1, \\ 0, & \text{if } \kappa > 1. \end{cases}$



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**The exclusion process on graphs with boundaries**

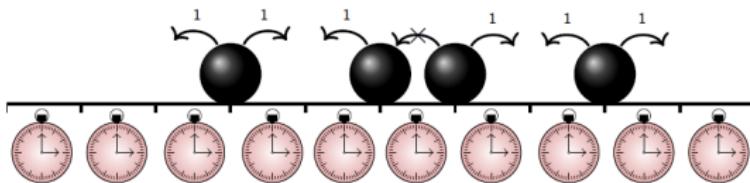
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## The exclusion process

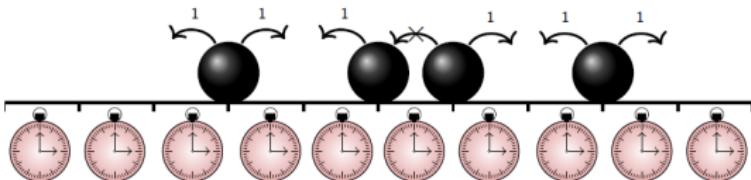
- *Informally:* Lay down  $k$  particles on a connected graph  $G = (V, E)$ . Let the particles move according to random walk rules, subject to the **exclusion rule**: no two particles may occupy the same vertex at any given time.



- This is a MC  $\{\eta_t : t \geq 0\}$  with state space  $\{0, 1\}^V$ . Total particle number is conserved.
- Any product Bernoulli measure  $\bigotimes_{x \in V} \text{Bern}(\alpha)$  of constant density  $\alpha \in [0, 1]$  is stationary for the process.

## The exclusion process

- *Informally:* Lay down  $k$  particles on a connected graph  $G = (V, E)$ . Let the particles move according to random walk rules, subject to the **exclusion rule**: no two particles may occupy the same vertex at any given time.



- **Scaling limit**

Take  $G_N = \{0, \frac{1}{N}, \frac{2}{N}, \dots, 1\}$  to be the discrete interval, and define the exclusion process on  $G_N$ . Accelerate the process by  $N^2$  (diffusive time scaling, RW  $\rightarrow$  BM).

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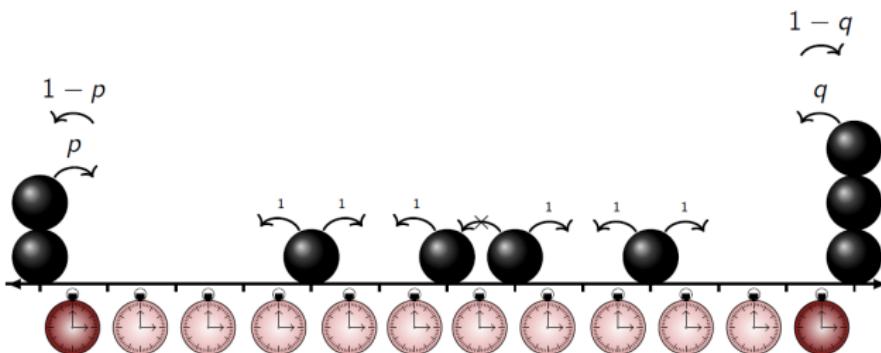
Exclusion  
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Exclusion cutoffs  
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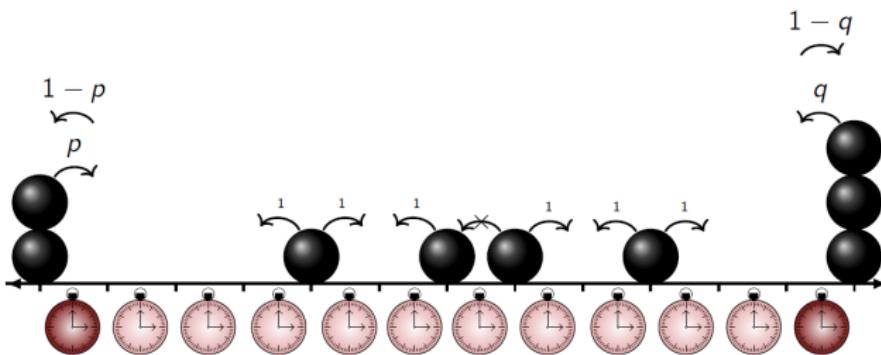
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Coda  
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## Bulk exclusion + boundary reservoirs



## Bulk exclusion + boundary reservoirs



## The cutoff problem for exclusion on the $d$ -dim lattice

- $d = 1$  Cutoffs have been established in several **exclusion** models and the related **adjacent transposition shuffle**.

[Wilson '04; Oliveira '13; Lacoin '16, '17; Labb  -Lacoin '19, '20; Nam-Nestoridi '19; Gantert-Nestoridi-Schmid '20]

Proofs take advantage of special properties of 1D particle systems: height function representations, stochastic monotonicity, censoring inequality, . . . .

- $d \geq 2$ : Lower bound on mixing time ✓; Upper bound? [off by  $2\times$ ]

Tools specific to  $d = 1$  do not generalize to higher dimensions.

## Good observables for exclusion processes

Let  $F : V_N \rightarrow \mathbb{R}$  be a test function (from a suitable Banach space).

**Empirical density**

$$\pi_t^N(F) = \frac{1}{|V_N|} \sum_{x \in V_N} \eta_{N^2 t}^N(x) F(x).$$

*Reason:* To obtain a **Law of Large Numbers (LLN)**.

$\pi_t^N \Rightarrow \pi_t$ , where  $\frac{d\pi_t}{dx} = \rho_t$  solves the heat equation  $\partial_t \rho_t = \Delta \rho_t$ .

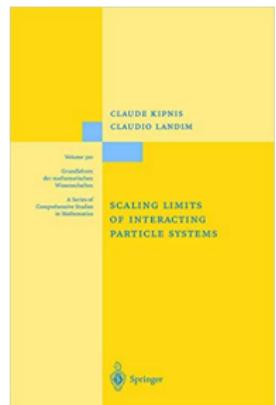
Mean density  $\rho_t^N(x) := \mathbb{E}_{\mu_N} [\eta_t^N(x)]$ .

**Density fluctuation field**

$$\mathcal{Y}_t^N(F) = \frac{1}{\sqrt{|V_N|}} \sum_{x \in V_N} \left( \eta_{N^2 t}^N(x) - \rho_{N^2 t}^N(x) \right) F(x).$$

*Reason:* To obtain a **Central Limit Theorem (CLT)**.

$\mathcal{Y}_t^N \Rightarrow \mathcal{Y}_t$ , an Ornstein-Uhlenbeck (Gaussian) process.



- To prove **scaling limits**, need to consider *all*  $F$  in the said Banach space.
- To prove **cutoff**, suffice to take  $F = \psi_1^N$ , the eigenfunction corresponding to the smallest nonzero eigenvalue of  $-\Delta_N$ .

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**Cutoffs for exclusion processes**

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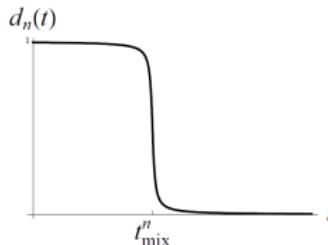
## Cutoffs for 1D symmetric exclusion

Discrete interval  $\{0, \frac{1}{N}, \frac{2}{N}, \dots, 1\}$

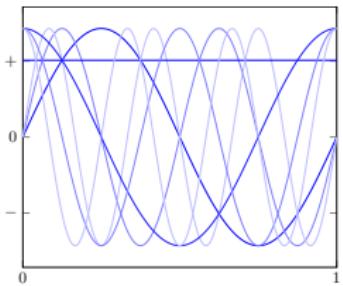
Endpoint boundary condition: Open (Dirichlet), Closed (Neumann), Periodic.

Boundary condition	Cutoff times $N^2 t_N$	#1 proof
● ●	$\frac{N^2 \log N}{2\pi^2}$	[Lacoin '16]
● ●	$\frac{N^2 \log N}{2(2\pi)^2}$	[Lacoin '16, '17]
● ●	$\frac{N^2 \log N}{2(\pi/2)^2}$	[Gantert–Nestoridi–Schmid '20]
● ●	$\frac{N^2 \log N}{2\pi^2}$	[C.–Jara–Marinho '20+]

Each purple value is the smallest nonzero eigenvalue of  $-u'' = \lambda u$   
on  $[0, 1]$  with the appropriate boundary condition:  $t_N = \frac{\log N}{2\lambda_1}$ .

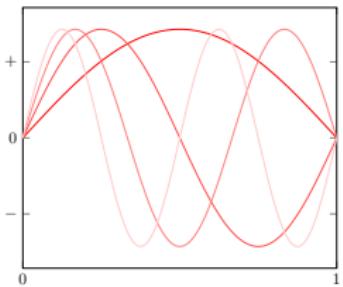


# 1D Laplacian eigenfunctions: Sines and cosines!

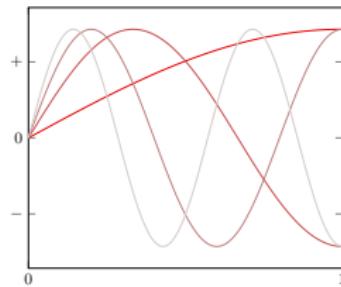


(a) Periodic b.c.

(b) Closed (Neumann) b.c.



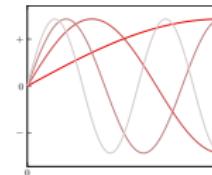
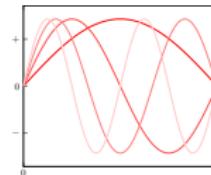
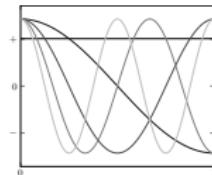
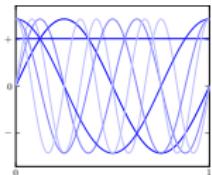
(c) Open (Dirichlet) b.c.



(d) Open at {0}, Closed at {1}

Taught in undergrad (P)DE courses! (I did this for an intro engineering DiffEq course at Cornell in 2010!)

# 1D continuous Laplacian eigensolutions

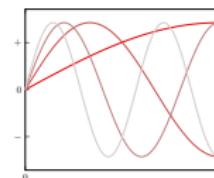
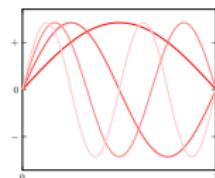
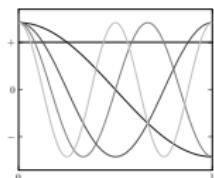
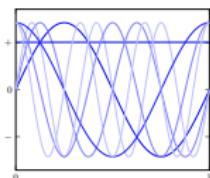


Unit interval  $[0, 1]$ ,  $-\Delta u := -u''$ .

Endpoint boundary condition: Open (Dirichlet), Closed (Neumann), Periodic.

Boundary condition	Eigenvalue $\lambda_k$	Eigenfunction $\psi_k(x)$	Index & range
● ●	$(2k\pi)^2$	$\cos(2k\pi x), \sin(2k\pi x)$	$k \in \mathbb{N} \cup \{0\}$
● ●	$(k\pi)^2$	$\cos(k\pi x)$	$k \in \mathbb{N} \cup \{0\}$
● ●	$(k\pi)^2$	$\sin(k\pi x)$	$k \in \mathbb{N}$
● ●	$(k\pi/2)^2$	$\sin\left(\frac{k\pi x}{2}\right)$	$k \in 2\mathbb{N} - 1$

# 1D discrete Laplacian eigensolutions



Discrete interval  $\{0, \frac{1}{N}, \frac{2}{N}, \dots, 1\}$ ,  $(-\Delta_N u)(x) := N^2 [2u(x) - u(x + \frac{1}{N}) - u(x - \frac{1}{N})]$ .  
 Endpoint boundary condition: Open (Dirichlet), Closed (Neumann), Periodic.

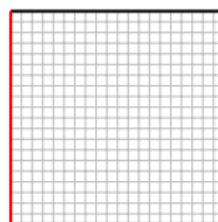
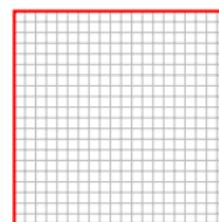
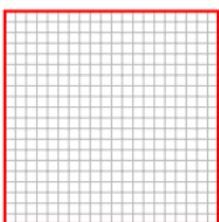
Boundary condition	Eigenvalue $\lambda_k^N$	Eigenfunction $\psi_k^N(x)$ $(x \in \{0, \frac{1}{N}, \frac{2}{N}, \dots, 1\})$	Index & range
●-----●	$2N^2 \left(1 - \cos\left(\frac{2k\pi}{N}\right)\right)$	$\cos(2k\pi x), \sin(2k\pi x)$	$k \in \{0, 1, \dots, N-1\}$ (Note: $\lambda_k^N = \lambda_{N-k}^N$ )
●-----●	$2N^2 \left(1 - \cos\left(\frac{k\pi}{N}\right)\right)$	$\cos(k\pi x)$	$k \in \{0, 1, \dots, N\}$
●-----●	$2N^2 \left(1 - \cos\left(\frac{k\pi}{N}\right)\right)$	$\sin(k\pi x)$	$k \in \{1, 2, \dots, N\}$
●-----●	$2N^2 \left(1 - \cos\left(\frac{k\pi}{2N}\right)\right)$	$\sin\left(\frac{k\pi x}{2}\right)$	$k \in \{1, 3, \dots, 2N-1\}$

**Spectral convergence:** For each  $k$ ,  $\lambda_k^N \xrightarrow[N \rightarrow \infty]{} \lambda_k$ , and  $\psi_k^N \xrightarrow[N \rightarrow \infty]{} \psi_k$  in  $C([0, 1])$ .

Cutoffs for  $d$ -dim symm. exclusion with open boundaries NEW!

Take  $I_1 \times I_2 \times \dots \times I_d$ , where each  $I_j \in \{\text{P, C, O, OC}\}$ . Use separation of variables!

$$\psi_k^{(N)}(x_1, \dots, x_d) = \prod_{j=1}^d \psi_{k_j}^{(N)}(x_j), \quad \lambda_k^{(N)} = \sum_{j=1}^d \lambda_{k_j}^{(N)}.$$



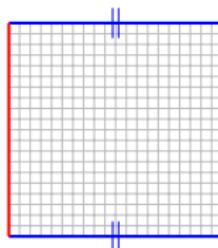
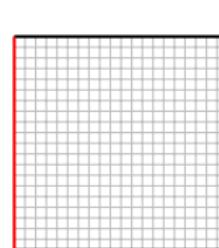
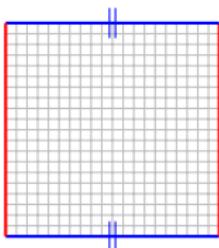
### Cutoffs established

[C.-Jara–Marinho '20+]

at times

$$\mathcal{T}_N t_N = \frac{N^2 \log(N^d)}{2\lambda_1}$$

assuming the cube has an  
open codim-1 boundary.



Examples for  $d = 2 \longrightarrow$

$$(d) \lambda_1 = \pi^2 + 0$$

$$(e) \lambda_1 = (\pi/2)^2 + 0$$

$$(f) \lambda_1 = (\pi/2)^2 + 0$$

# Cutoffs for symm. exclusion on graphs with open boundaries

## Basic idea

Let  $(K, d, \mathfrak{m})$  be a compact metric measure space with boundary  $\partial K$  equipped with a normalized surface measure  $\mathfrak{s}$ .

Approximate  $(K, d, \mathfrak{m}, \partial K, \mathfrak{s})$  by an increasing family of bounded-degree graphs  $\{G_N = (V_N, E_N)\}_{N \in \mathbb{N}}$ , each having a boundary vertex set  $\partial V_N \subset V_N$ .

### Assumption 1 (Geometric convergence).

(a) For every  $N \in \mathbb{N}$ ,  $V_N \subseteq K$  and  $\partial V_N \subseteq \partial K$ .

Moreover, as  $N \rightarrow \infty$ :

(b)  $|\partial V_N|/|V_N| \rightarrow 0$ .

(c)  $\mathfrak{m}_N := \frac{1}{|V_N|} \sum_{x \in V_N} \delta_x$  converges weakly to  $\mathfrak{m}$ .

(d)  $\mathfrak{s}_N := \frac{1}{|\partial V_N|} \sum_{a \in \partial V_N} \delta_a$  converges weakly to  $\mathfrak{s}$ .

## Cutoffs for symm. exclusion on graphs with open boundaries

### Exclusion process induced Laplacian

$\{\mathcal{T}_N\}_N$  is the **diffusive time scale**  $\uparrow \infty$ .

$$(\Delta_N f)(x) = \mathcal{T}_N \sum_{\substack{y \in V_N \\ y \sim x}} (f(y) - f(x)) - \mathcal{T}_N \lambda_{\Sigma, N}(x) f(x) \mathbb{1}_{\{x \in \partial V_N\}}, \quad x \in V_N.$$

The operator  $-\Delta_N$  is self-adjoint on  $L^2(\mathfrak{m}_N)$ . Denote the  $L^2(\mathfrak{m}_N)$ -orthonormal eigensolutions of  $-\Delta_N$  by  $\{(\lambda_i^N, \psi_i^N)\}_i$ , where  $0 < \lambda_1^N < \lambda_2^N \leq \lambda_3^N \leq \dots$

#### Assumption 2 (Spectral convergence).

- (a) There exist  $\{\lambda_i\}_{i=1}^\infty \subseteq \mathbb{R}_+$  such that  $\lim_{N \rightarrow \infty} \lambda_i^N = \lambda_i$  for each  $i \in \mathbb{N}$ .
- (b) There exist  $\{\psi_i\}_{i=1}^\infty \subseteq C(K)$  such that  $\psi_i^N \rightarrow \psi_i$  in  $C(K)$  for each  $i \in \mathbb{N}$ .
- (c)  $\psi_1$  is Lipschitz continuous on  $K$ .
- (d) Each  $\psi_i$  has bounded normal derivative on  $\partial K$ .

## Cutoffs for symm. exclusion on graphs with open boundaries

### Exclusion process induced Laplacian

$\{\mathcal{T}_N\}_N$  is the diffusive time scale  $\uparrow \infty$ .

$$(\Delta_N f)(x) = \mathcal{T}_N \sum_{\substack{y \in V_N \\ y \sim x}} (f(y) - f(x)) - \mathcal{T}_N \lambda_{\Sigma, N}(x) f(x) \mathbb{1}_{\{x \in \partial V_N\}}, \quad x \in V_N.$$

The operator  $-\Delta_N$  is self-adjoint on  $L^2(\mathfrak{m}_N)$ . Denote the  $L^2(\mathfrak{m}_N)$ -orthonormal eigensolutions of  $-\Delta_N$  by  $\{(\lambda_i^N, \psi_i^N)\}_i$ , where  $0 < \lambda_1^N < \lambda_2^N \leq \lambda_3^N \leq \dots$

$$t_N := \frac{\log |V_N|}{2\lambda_1^N},$$

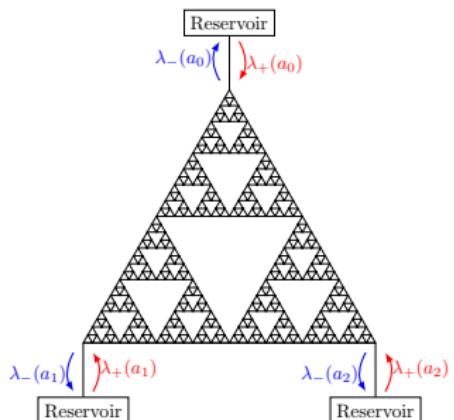
Set

$$\kappa^* := \sup \left\{ \kappa > 0 : \varlimsup_{N \rightarrow \infty} \frac{|V_N|^\kappa}{\mathcal{T}_N |\partial V_N|} \sum_{i=1}^{|V_N|} \frac{|\int \psi_i^N d\mathfrak{m}_N|}{\lambda_i^N} = 0 \right\}.$$

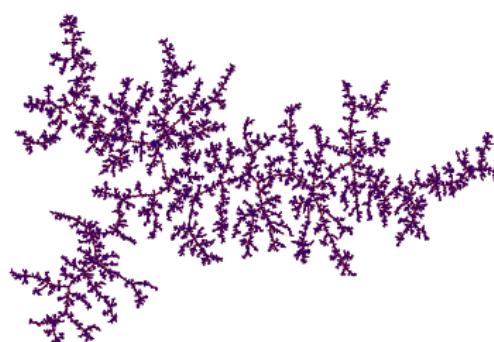
### Cutoff Theorem [C.-Jara–Marinho '20+].

Suppose  $\kappa^* > 1$ . Then for every  $\varepsilon \in (0, 1)$ ,  $\lim_{N \rightarrow \infty} \frac{t_{\text{mix}}^N(\varepsilon)}{t_N \mathcal{T}_N} = 1$ .

## Cutoff examples: non-lattice graphs NEW!



(a) Sierpinski gasket



(b) Aldous' continuum random tree

Geometric convergence ✓

Spectral convergence ✓ [Barlow–Perkins '89, Kigami '90s, ...; Croydon–Hambly '08]

Cutoff can also be established on  $d$ -fold product of these graphs.

Motivations  
ooo

Convergence  
ooo

Cutoff  
ooo

Exclusion  
oooo

Exclusion cutoffs  
oooooooo

Proof ideas  
●ooo

Coda  
oo

# Outline

Motivations

Convergence of Markov chains

The cutoff phenomenon

The exclusion process on graphs with boundaries

Cutoffs for exclusion processes

## Proof ideas

Thank you for listening!

# Coupling

**Couple** the following three MCs:

- $\eta_t^{1,N}$  - original exclusion process, initialized from  $\eta$ .  
*Boundary rate:*  $\lambda_{+,N}(a)$  [in],  $\lambda_{-,N}(a)$  [out].
- $\eta_t^{2,N}$  - indep. copy of  $\eta_t^{1,N}$ , initialized from  $\eta'$ .  
*Boundary rate:*  $\lambda_{+,N}(a)$  [in],  $\lambda_{-,N}(a)$  [out].
- $\xi_t^N$  - indep. copy of  $\eta_t^{1,N}$ , initialized from  $\eta$ .  
*Boundary rate:* 0 [in],  $\lambda_{+,N}(a) + \lambda_{-,N}(a)$  [out]. (**pure annihilation**)

→ Same bulk dynamics; boundary dynamics differ between  $\{\eta_t^{1,N}, \eta_t^{2,N}\}$  and  $\xi_t^N$ .

Let

- $\eta_{\text{full}}$  denote the all-1 config:  $\eta_{\text{full}} \equiv 1$ . Initialize from the fully occupied config.
- $T_N := \inf\{t \geq 0 : \xi_t^N(x) = 0 \quad \forall x \in V_N\}$ . #1 time the graph is empty of particles.

**Prop.**  $d_N(t) = \mathbb{P}_{\eta_{\text{full}}} [T_N > t]$ .

From this point on, the analysis is performed on this **worst-case model**.

## Lower bound on mixing times: Wilson's method

Well-known method from [Wilson '04].

- Use the empirical density observable

$$\pi_t^N(\psi_1^N) = \frac{1}{|V_N|} \sum_{x \in V_N} \eta_{T_N t}^N(x) \psi_1^N(x).$$

- Apply Dynkin's formula:  $\exists$  martingale  $\{\mathcal{M}_t^N : t \geq 0\}$  s.t.

$$\pi_t^N(\psi_1^N) = \pi_0^N(\psi_1^N) e^{-\lambda_1^N t} + \mathcal{M}_t^N.$$

- Compute the mean and the variance of  $\pi_t^N(\psi_1^N)$ , deduce that there exists  $c > 0$  s.t. for every  $t \geq 0$  and every  $N \in \mathbb{N}$ ,

$$d_N(t T_N) \geq 1 - \frac{c e^{2\lambda_1^N t}}{|V_N|}.$$

This implies that for any  $\kappa < 1$ ,  $d_N(\kappa t_N T_N) \xrightarrow[N \rightarrow \infty]{} 1$  (desired lower bound).

## Upper bound on mixing times: the “cutoff martingale” method

NEW!

- Use the density fluctuation field observable, suitably rescaled:  
**(mean zero) Cutoff martingale**

$$\mathcal{X}_\kappa^N(\psi_1^N) = |V_N|^{\frac{\kappa}{2}-1} \sum_{x \in V_N} (\eta_{\kappa t_N \mathcal{T}_N}^N(x) - \rho_{\kappa t_N \mathcal{T}_N}^N(x)) \psi_1^N(x) \quad \left[ t_N = \frac{\log |V_N|}{2\lambda_1^N} \right]$$

- On the one hand, by expanding  $\rho_t^N$  in a spectral series, we find that w.p. 1,  $\forall \kappa \geq 0$ ,

$$\mathcal{X}_\kappa^N(\psi_1^N) = |V_N|^{\frac{\kappa}{2}-1} \sum_{x \in V_N} \eta_{\kappa t_N \mathcal{T}_N}^N(x) \psi_1^N(x) - \int \psi_1^N d\mathfrak{m}_N.$$

If  $\kappa t_N \mathcal{T}_N$  is the time that all particles are annihilated (reach stationarity), then  $\chi_\kappa^N(\psi_1^N) = \Theta_N(1)$  w.p. 1.

- On the other hand, a **laborious computation** shows that for every  $\kappa > 1$ ,

$$\lim_{N \rightarrow \infty} \mathbb{E}_{\mu_{\text{full}}} [\chi_\kappa^N(\psi_1^N)]^2 = \infty$$

**Laborious computation** = analysis of the mean quadratic variation ( $\mathbb{E}\text{QV}$ ) of the cutoff MG.  
 $\mathbb{E}\text{QV}$  is expanded in a **complete Fourier series**, then simplified using **Dirichlet forms** and  $\Gamma$ -calculus.  
We also need the **negativity of the two-point correlation function in the exclusion process**.

- So all particles must have been annihilated by time  $t_N \mathcal{T}_N$ . i.e., for any  $\kappa > 1$ ,  
 $d_N(\kappa t_N \mathcal{T}_N) \xrightarrow[N \rightarrow \infty]{} 0$  (desired upper bound). Q.E.D.

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## Cutoffs for exclusion with closed/periodic boundaries

[C.-Marinho '20+]

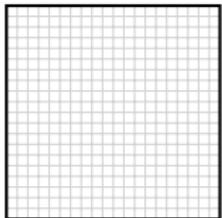
Adapting the cutoff MG method to prove cutoffs for graphs with closed/no boundaries.

Works in any dimension  $d$ .

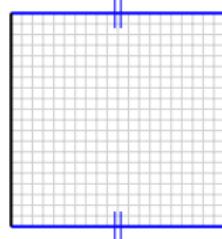
- A revised coupling argument.
- A revised cutoff MG which removes the nonzero stationary density, and therefore captures the fluctuations about stationarity.
- Analysis of the cutoff MG is similar to the open boundary case.

Cutoffs at times

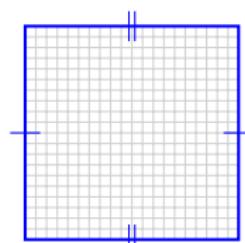
$$\mathcal{T}_N t_N = \frac{N^2 \log(N^d)}{2\lambda_1}$$



(g)  $\lambda_1 = \pi^2 + 0$



(h)  $\lambda_1 = \pi^2 + 0$



(i)  $\lambda_1 = (2\pi)^2 + 0$

Examples for  $d = 2 \longrightarrow$

**Open questions:** Can the cutoff MG method be applied to prove the **cutoff window**? What about **other prob. models** on graphs, such as **asymmetric exclusion processes**?