

# Convergence in High Probability of the Quantum Diffusion in a Random Band Matrix Model

Project supervised by Prof. Antti Knowles- ETH Zürich

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# Introduction of the model

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- Let  $\mathbb{Z}^d$  be the infinite lattice with the Euclidean norm  $|\cdot|_{\mathbb{Z}^d}$  and let  $M$  the number of points situated at distance at most  $W$  ( $W \geq 2$ ) from the origin, i.e.

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- For simplicity we consider through the proof a  $d$ -dimensional finite periodic lattice  $\Lambda_N \subset \mathbb{Z}^d$  ( $d \geq 1$ ) of linear size  $N$  equipped with the Euclidean norm  $|\cdot|_{\mathbb{Z}^d}$ . Specifically, we take  $\Lambda_N$  to be a cube centered around the origin with side length  $N$ , i.e.

$$\Lambda_N := ([-N/2, N/2) \cap \mathbb{Z})^d.$$

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We define the *random band matrix*  $(H_{xy})$  through

$$H_{xy} := \sqrt{S_{xy}} A_{xy},$$

where  $(A_{xy})$  is Hermitian random matrix whose upper triangular entries  $(A_{xy} : x \leq y)$  are independent random variables uniformly distributed on the unit circle  $\mathbb{S}^1 \subset \mathbb{C}$ .

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- Note that the entries  $H_{xy}$  in the random band matrix  $H$  are indexed by  $x$  and  $y$  which are indices of points in  $\Lambda_N$ .



# Introduction of the Model

- Let us consider also the function

$$P(t, x) = |(e^{-itH/2})_{0x}|^2,$$

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- For  $\kappa > 0$ , we introduce the macroscopic time and space coordinates  $T$  and  $X$ , which are independent of  $W$ , and consider the microscopic time and space coordinates

$$t = W^{d\kappa} T,$$

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$$\begin{aligned}t &= W^{d\kappa} T, \\x &= W^{1+d\kappa/2} X.\end{aligned}$$

- Given  $\phi \in C_b(\mathbb{R}^d)$ , we define the main quantity that we investigate by

$$Y_{T,\kappa,W}(\phi) \equiv Y_T(\phi) := \sum_x P(W^{d\kappa} T, x) \phi\left(\frac{x}{W^{1+d\kappa/2}}\right).$$

# Known and new result

## Theorem (Erdős L., Knowles A. 2011)

Let  $0 < \kappa < 1/3$  be fixed. Then for any  $\phi \in C_b(\mathbb{R}^d)$  and for any  $T_0 > 0$  we have that

$$\lim_{W \rightarrow \infty} \mathbb{E} Y_T(\phi) = \int_{\mathbb{R}^d} dX L(T, X) \phi(X),$$

uniformly in  $N \geq W^{1+d/6}$  and  $0 \leq T \leq T_0$ .

Here

$$L(T, X) := \int_0^1 d\lambda \frac{4}{\pi} \frac{\lambda^2}{\sqrt{1-\lambda^2}} G(\lambda T, X),$$

and  $G$  is the heat kernel

$$G(T, X) := \left( \frac{d+2}{2\pi T} \right)^{d/2} e^{-\frac{d+2}{2T} |X|^2}.$$

## Theorem (Knowles A., M. 2015)

Fix  $T_0 > 0$  and  $\kappa$  such that  $0 < \kappa < 1/3$ . Choose a real number  $\beta$  satisfying  $0 < \beta < 2/3 - 2\kappa$ . Then there exists  $C \geq 0$  and  $W_0 \geq 0$  depending only on  $T_0$ ,  $\kappa$  and  $\beta$  such that for all  $T \in [0, T_0]$ ,  $W \geq W_0$  and  $N \geq W^{1+\frac{d}{6}}$  we have

$$\text{Var}(Y_T(\phi)) \leq \frac{C \|\phi\|_\infty^2}{W^{d\beta}}.$$

# Sketch of the Proof

Expanding in non-backtracking powers

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Let us introduce

$$H^{(1)} = H,$$

$$H_{x_0, x_n}^{(n)} := \sum_{x_1, \dots, x_{n-1}} \left( \prod_{i=0}^{n-2} \mathbf{1}(x_i \neq x_{i+2}) \right) H_{x_0 x_1, \dots, x_{n-1} x_n} \quad (n \geq 2).$$



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Let  $U_k$  be the  $k$ -th Chebyshev polynomial of the second kind and let

$$\alpha_k(t) := \frac{2}{\pi} \int_{-1}^1 \sqrt{1 - \zeta^2} e^{-it\zeta} U_k(\zeta) d\zeta.$$

We define the quantity  $a_m(t) = \sum_{k \geq 0} \frac{\alpha_{m+2k}(t)}{(M-1)^k}$ . Then we have that

$$e^{-itH/2} = \sum_{m \geq 0} a_m(t) H^{(m)}.$$

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Plugging in the definition of  $Y_T(\phi)$  we have

$$\begin{aligned} \text{Var}(Y_T(\phi)) &= \sum_{y_1, y_2} \phi\left(\frac{y_1}{W^{1+d\kappa/2}}\right) \phi\left(\frac{y_2}{W^{1+d\kappa/2}}\right) \langle P(t, y_1); P(t, y_2) \rangle \\ &\leq \|\phi\|_\infty^2 \sum_{y_1} \sum_{y_2} |\langle P(t, y_1); P(t, y_2) \rangle|. \end{aligned}$$

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Moreover,

$$\begin{aligned} \langle P(t, y_1); P(t, y_2) \rangle &= \\ &= \sum_{n_{11}, n_{12} \geq 0} \sum_{n_{21}, n_{22} \geq 0} a_{n_{11}}(t) \overline{a_{n_{12}}(t)} a_{n_{21}}(t) \overline{a_{n_{22}}(t)} \langle H_{0y_1}^{(n_{11})} H_{y_1 0}^{(n_{12})}; H_{0y_2}^{(n_{21})} H_{y_2 0}^{(n_{22})} \rangle. \end{aligned}$$

# Graphical Representation

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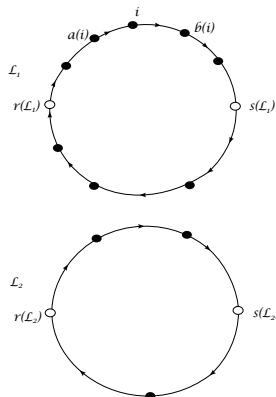


Figure: The graphical representation of a path of vertices.

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- For each configuration of labels  $\mathbf{x}$  we assign a *lumping*  $\Gamma = \Gamma(\mathbf{x})$  of the set of edges  $E(\mathcal{L})$ . The lumping  $\Gamma = \Gamma(\mathbf{x})$  associated with the labels  $\mathbf{x}$  is given by the equivalence relation

$$e \sim e' \Leftrightarrow \{x_{a(e)}, x_{b(e)}\} = \{x_{a(e')}, x_{b(e')}\}.$$



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- Using the graph  $\mathcal{L}$  we may now write the covariance as

$$\langle H_{0y_1}^{(n_{11})} H_{y_1 0}^{(n_{12})}; H_{0y_2}^{(n_{21})} H_{y_2 0}^{(n_{22})} \rangle = \sum_{\mathbf{x} \in \Lambda_N^{V(\mathcal{L})}} Q_{y_1, y_2}(\mathbf{x}) A(\mathbf{x}),$$

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where

$$Q_{y_1, y_2}(\mathbf{x}) = \mathbf{1}(x_{r(\mathcal{L}_1)} = 0) \mathbf{1}(x_{r(\mathcal{L}_2)} = 0) \mathbf{1}(x_{s(\mathcal{L}_1)} = y_1) \mathbf{1}(x_{s(\mathcal{L}_2)} = y_2) \prod_{i \in V_b(\mathcal{L})} \mathbf{1}(x_{a(i)} \neq x_{b(i)}).$$

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and

$$A(\mathbf{x}) = \mathbb{E} \prod_{e \in E(\mathcal{L})} H_{x_e} - \mathbb{E} \prod_{e \in E(\mathcal{L}_1)} H_{x_e} \mathbb{E} \prod_{e \in E(\mathcal{L}_2)} H_{x_e}.$$

# Graphical Representation

- Let  $\mathfrak{B}_c(E(\mathcal{L}))$  be the set of connected even lumpings, i.e. the set of all lumpings  $\Gamma$  for which each lump  $\gamma \in \Gamma$  has even size and there exists  $\gamma \in \Gamma$  such that  $\gamma \cap E(\mathcal{L}_k) \neq \emptyset$ , for  $k \in \{1, 2\}$ .

- Let  $\mathfrak{P}_c(E(\mathcal{L}))$  be the set of connected even lumpings, i.e. the set of all lumpings  $\Gamma$  for which each lump  $\gamma \in \Gamma$  has even size and there exists  $\gamma \in \Gamma$  such that  $\gamma \cap E(\mathcal{L}_k) \neq \emptyset$ , for  $k \in \{1, 2\}$ .

## Lemma

*We have that*

$$\langle H_{0y_1}^{(n_{11})} H_{y_1 0}^{(n_{12})}; H_{0y_2}^{(n_{21})} H_{y_2 0}^{(n_{22})} \rangle = \sum_{\Gamma \in \mathfrak{P}_c(E(\mathcal{L}))} V_{y_1, y_2}(\Gamma),$$

where  $V_{y_1, y_2}(\Gamma) = \sum_{\mathbf{x}} \mathbf{1}(\Gamma(\mathbf{x}) = \Gamma) Q_{y_1, y_2}(\mathbf{x}) A(\mathbf{x})$ .

# Graphical Representation



- We define  $\mathfrak{M}_c$  the set of all connected pairings

$$\bigsqcup_{n_{11}, n_{12}, n_{21}, n_{22}} \{ \Pi \in \mathfrak{P}_c(E(\mathcal{L}(n_{11}, n_{12}, n_{21}, n_{22}))) : |\pi| = 2, \forall \pi \in \Pi \}.$$

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- We call the lumps  $\pi \in \Pi$  of a pairing  $\Pi$  *bridges*.

# First estimate

- Consider

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We have

$$\begin{aligned} & |\langle P(t, y_1); P(t, y_2) \rangle| \\ & \leq \sum_{\Pi \in \mathfrak{M}_c} |a_{n_{11}(\Pi)}(t) \overline{a_{n_{12}(\Pi)}(t)} a_{n_{21}(\Pi)}(t) \overline{a_{n_{22}(\Pi)}(t)}| \\ & \sum_{\mathbf{x}} Q_{y_1, y_2}(\mathbf{x}) \prod_{\{e, e'\} \in \Pi} S_{x_e} \prod_{\{e, e'\} \in \Pi} J_{\{e, e'\}}(\mathbf{x}). \end{aligned}$$

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- To each  $\Pi \in \mathfrak{M}_c$  we associate a couple  $(\Sigma, l_\Sigma)$ , where  $\Sigma \in \mathfrak{M}_c$  has no parallel bridges and  $l_\Sigma := (l_\sigma)_{\sigma \in \Sigma} \in \mathbb{N}^\Sigma$ . The integer  $l_\sigma$  denotes the number of parallel bridges of  $\Pi$  that were collapsed into the bridge  $\sigma$  of  $\Sigma$ . Inverting the procedure we obtain a bijection  $\Pi \longleftrightarrow (\Sigma, l_\Sigma)$ .

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- We further define the set of admissible skeletons as

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- We further define  $|l_\Sigma| = \sum_{\sigma \in \Sigma} l_\sigma$  for  $\Sigma \in \mathfrak{G}$ .

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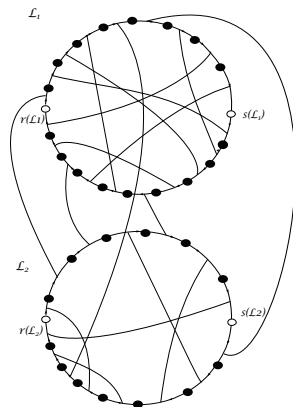


Figure: Graphical representation of the skeleton for a given configuration

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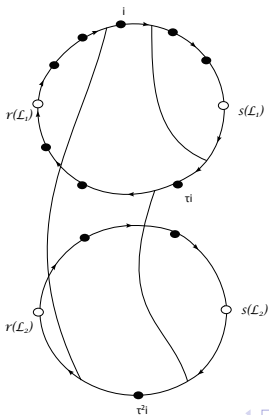
where

$$R(\Sigma) = \sum_{\mathbf{x} \in \Lambda_N^{V(\Sigma)}} \mathbf{1}(x_{r(\mathcal{L}_1(\Sigma))} = 0) \mathbf{1}(x_{r(\mathcal{L}_2(\Sigma))} = 0) \prod_{\{e, e'\} \in \Sigma} \left( S^{I_{\{e, e'\}}} \right)_{x_e} \prod_{\sigma \in \Sigma} J_\sigma(\mathbf{x}).$$

# Orbits of vertices

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- For fixed  $\Sigma \in \mathfrak{G}$  we define  $\tau : V(\Sigma) \rightarrow V(\Sigma)$  as follows. Let  $i \in V(\Sigma)$  and let  $e$  be the unique edge such that  $\{\{i, b(i)\}, e\} \in \Sigma$ . Then, for any vertex  $i$  of  $\Sigma \in \mathfrak{G}$  we define  $\tau i = b(e)$ . We denote the orbit of the vertex  $i \in \Sigma$  by  $[i] := \{\tau^n i : n \in \mathbb{N}\}$



# Orbits of vertices

- Let  $Z(\Sigma) := \{[i] : i \in V(\Sigma)\}$  be the set of orbits of  $\Sigma$  and  $|\Sigma|$  be the number of bridges of the skeleton  $\Sigma$  and let  $L(\Sigma) = |Z^*(\Sigma)|$  with  $Z^*(\Sigma) := Z(\Sigma) \setminus \{[r(\mathcal{L}_1)], [r(\mathcal{L}_2)]\}$ .



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## Lemma

Let  $\Sigma \in \mathfrak{G}$  and  $l_\Sigma \in \Lambda_N^{V(\Sigma)}$ . We have that

$$R(\Sigma) \leq C \left( \frac{M}{M-1} \right)^{|l_\Sigma|} M^{-|\Sigma|/3+2/3}.$$

# Truncation

- We introduce a cut-off at  $|I_\Sigma| < M^\mu$  for  $\mu < 1/3$ . We define

$$E^\leq = \sum_{\Sigma \in \mathcal{G}} \sum_{|I_\Sigma| \leq M^\mu} |a_{n_{11}}(\Sigma, I_\Sigma)(t) \overline{a_{n_{12}}(\Sigma, I_\Sigma)(t)} a_{n_{21}}(\Sigma, I_\Sigma)(t) \overline{a_{n_{22}}(\Sigma, I_\Sigma)(t)}| R(\Sigma)$$

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## Lemma

- 1 For any time  $t$  and for any  $n \in \mathbb{N}$  we have  $|a_n(t)| \leq \frac{Ct^n}{n!}$ , with  $C$  universal constant.
- 2 We have  $\sum_{n \geq 0} |a_n(t)|^2 = 1 + O(M^{-1})$ , uniformly in  $t \in \mathbb{R}$ .

# Main Lemma

## Lemma

For any  $\Sigma \in \mathfrak{G}$  with  $|\Sigma| \geq 3$  we have

$$\sum_{I_\Sigma} \mathbf{1}(|I_\Sigma| \leq M^\mu) |a_{n_{11}(\Sigma, I_\Sigma)}(t) \overline{a_{n_{12}(\Sigma, I_\Sigma)}(t)} a_{n_{21}(\Sigma, I_\Sigma)}(t) \overline{a_{n_{22}(\Sigma, I_\Sigma)}(t)}| \\ \leq \frac{CM^{\mu(|\Sigma|-2)}}{(|\Sigma| - 3)!}.$$

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In order to finish the argument

- 1 We prove using Stirling approximation and some arguments involving geometric series that the remaining terms are tiny.

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- 2 We find bounds by direct computation for the cases  $|\Sigma| = 0$ ,  $|\Sigma| = 1$  and  $|\Sigma| = 2$ , and we conclude the proof.

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- Truncating the expression in  $|I_\Sigma|$  and summing the two terms under the specific conditions on the parameters imposed by the setting (i.e.  $0 < \kappa \leq 1/3$ ,  $0 < \beta < 2/3 - 2\kappa$ ), we obtain a bound for the variance of the form

$$\text{Var}(Y_T(\phi)) \leq \frac{C \|\phi\|_\infty^2}{W d^\beta}.$$

**Thank you for your attention!**

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