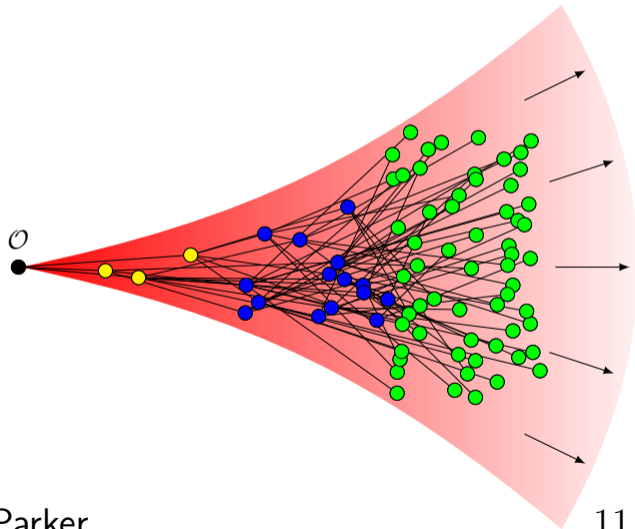
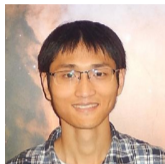

A Universal Operator Growth Hypothesis



Acknowledgements

Collaborators



Xiangyu Cao



Ehud Altman



Thomas Scaffidi



Alex Avdoshkin
Aavishkar Patel
Jaewon Kim

Funding



Advisor



Joel Moore

Quantum Mechanics

Microscopic description of the system.

Example: Chaotic Ising Model

$$H = \sum_i X_i + 1.05 Z_i Z_{i+1} + 0.59 Z_i$$

Correlation functions:

$$C(t) = \langle \mathcal{O}(t, x) \mathcal{O}(0) \rangle$$

Hard Solution: Hamiltonian dynamics

$$\mathcal{O}(t) = e^{-iHt} \mathcal{O} e^{iHt}.$$

Exact and **reversible** dynamics.

Hydrodynamics

Macroscopic description of quantum systems as classical PDEs.

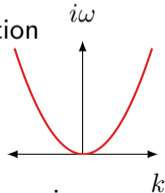
Example: Diffusion of energy

$$\frac{\partial}{\partial t} \varepsilon(t, x) = D \nabla^2 \varepsilon(t, x) + \nabla f,$$

with D diffusion, f thermal noise.

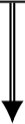
Easy Solution: Green's function

$$G(i\omega, k) = \frac{1}{i\omega + Dq^2}$$



Approximate & **irreversible** dynamics.

Quantum Mechanics



Operator
Growth
Hypothesis



Thermalization



Chaos



Complexity



Hydrodynamics

Operator Growth

or

How I learned to stop worrying and love tridiagonalization.

Operator Space

Inspiration: random unitary circuits.

- ▶ Keyserlingk, Rakovszky, Pollmann, Sondhi, 2017; Nahum, Vijay, Haah, 2017.
- ▶ Khemani, Vishwanath, Huse, 2017.

Consider a spin-1/2 system in d -dimensions with translation invariance.

$$H = \sum_{x \in \mathbb{Z}^d} h_x.$$

We abstract to the **space of operators**.

operators are “rounded” kets $|\mathcal{O}\rangle$

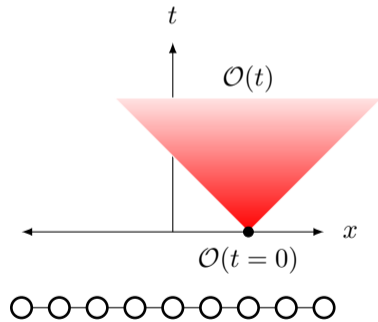
an example is $|\mathcal{O}\rangle = X_1 \otimes Y_2 \otimes Z_3 + 0.3Y_1 \otimes X_2$

the inner product is $\langle A|B\rangle := \text{Tr}[A^\dagger B] / \text{Tr}[1]$

the Liouvillian generalizes the Hamiltonian $\mathcal{L} = [H, \cdot]$.

time-evolution from Heisenberg EOM $-i \frac{d|\mathcal{O}\rangle}{dt} = \mathcal{L}|\mathcal{O}\rangle$.

solution $|\mathcal{O}(t)\rangle = e^{i\mathcal{L}t} |\mathcal{O}\rangle$



Three Observables

A. Correlation Function

$$C(t) := (\mathcal{O}(t)|\mathcal{O}(0)) = \sum_{n=0}^{\infty} \frac{\mu_{2n}}{(2n)!} (it)^{2n} \quad \text{with **moments** } \mu_{2n} = (\mathcal{O}|\mathcal{L}^{2n}|\mathcal{O}).$$

B. Green's Function

$$G(z) := (\mathcal{O}|\frac{1}{z - \mathcal{L}}|\mathcal{O}) = i \int_0^{\infty} e^{-izt} C(t) dt = \sum_{n=0}^{\infty} \frac{\mu_{2n}}{z^{2n+1}}$$

C. Spectral Function

$$\Phi(\omega) = \int_{-\infty}^{\infty} e^{-i\omega t} C(t) dt = \sum_{E, E'} |\langle E|\mathcal{O}|E'\rangle|^2 \delta(\omega - (E - E')).$$

The Graph of Operators

Example: Chaotic Ising Model

$$H = \sum_i X_i + 1.05Z_i Z_{i+1} + 0.5Z_i.$$

Problem: Compute $C(t) = (\mathcal{O}|e^{i\mathcal{L}t}|\mathcal{O})$.

$$\mathcal{O}(t) = e^{i\mathcal{L}t}\mathcal{O} = \mathcal{O} + (it)\mathcal{L}\mathcal{O} + (it)^2\mathcal{L}^2\mathcal{O} + \dots$$

The Graph of Operators

Example: Chaotic Ising Model

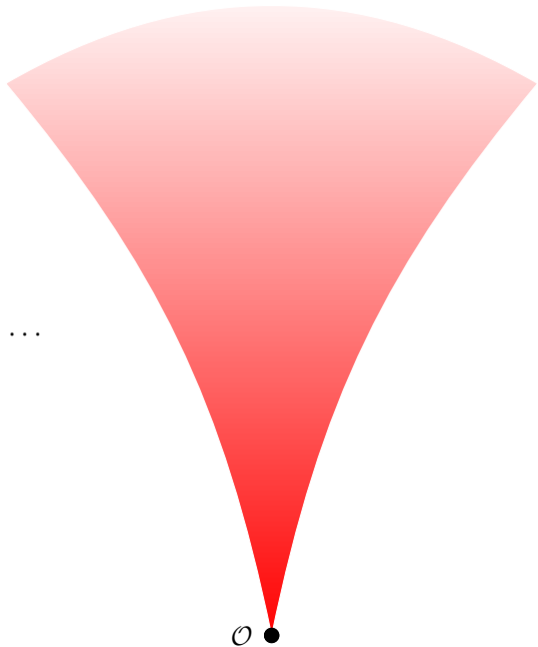
$$H = \sum_i X_i + 1.05Z_i Z_{i+1} + 0.5Z_i.$$

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$$\mathcal{O}(t) = e^{i\mathcal{L}t}\mathcal{O} = \mathcal{O} + (it)\mathcal{L}\mathcal{O} + (it)^2\mathcal{L}^2\mathcal{O} + \dots$$

Let's compute!

$$\mathcal{O} = X_1$$



The Graph of Operators

Example: Chaotic Ising Model

$$H = \sum_i X_i + 1.05Z_i Z_{i+1} + 0.5Z_i.$$

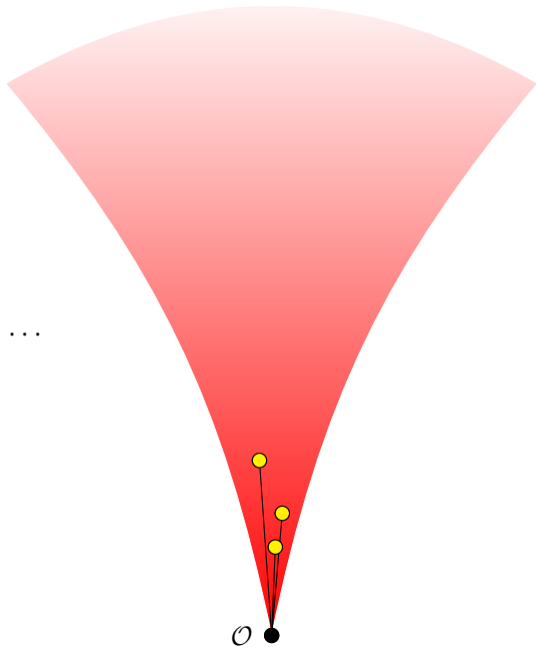
Problem: Compute $C(t) = (\mathcal{O} | e^{i\mathcal{L}t} | \mathcal{O})$.

$$\mathcal{O}(t) = e^{i\mathcal{L}t} \mathcal{O} = \mathcal{O} + (it)\mathcal{L}\mathcal{O} + (it)^2\mathcal{L}^2\mathcal{O} + \dots$$

Let's compute!

$$\mathcal{O} = X_1$$

$$\mathcal{L}\mathcal{O} = 1.05iY_1Z_2 + 1.05iZ_1Y_2 + 0.5iY_1$$



The Graph of Operators

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$$H = \sum_i X_i + 1.05Z_i Z_{i+1} + 0.5Z_i.$$

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Let's compute!

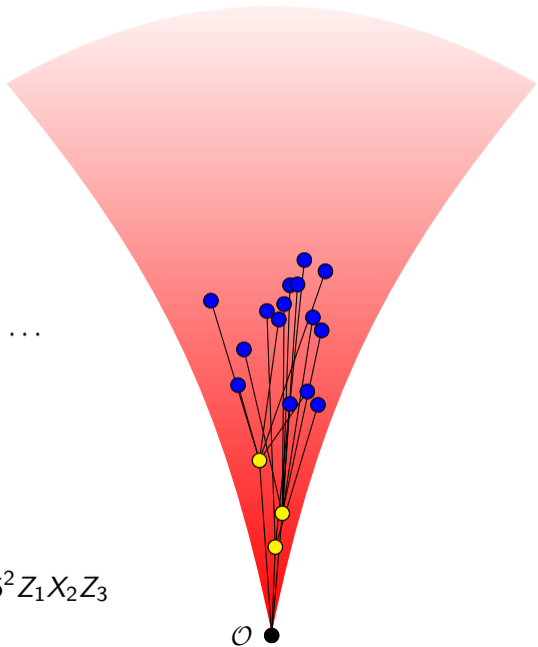
$$\mathcal{O} = X_1$$

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$$\mathcal{L}^2\mathcal{O} = 2.1Z_1Z_2 - 2.1Y_1Y_2$$

$$+ 1.05^2 Z_0 X_1 Z_2 + 1.05^2 X_1 + 1.05^2 X_2 + 1.05^2 Z_1 X_2 Z_3$$

$$+ 0.525 X_1 Z_2 + 0.525 Z_1 X_2 + 0.25 X_1.$$



The Graph of Operators

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Let's compute!

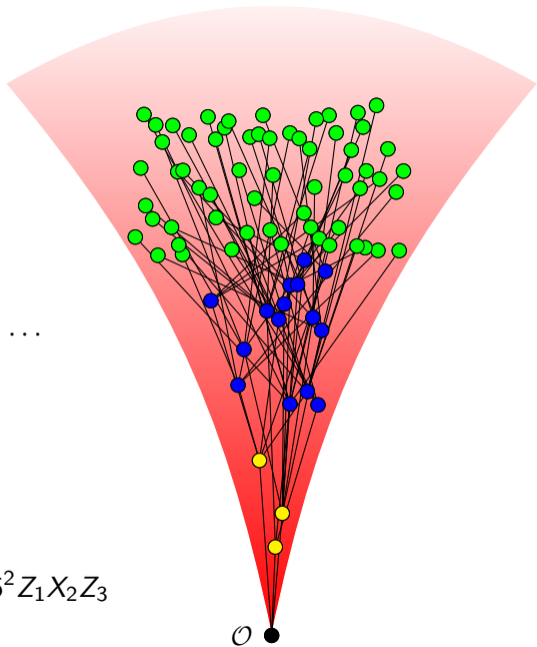
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The Graph of Operators

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$$H = \sum_i X_i + 1.05Z_i Z_{i+1} + 0.5Z_i.$$

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Let's compute!

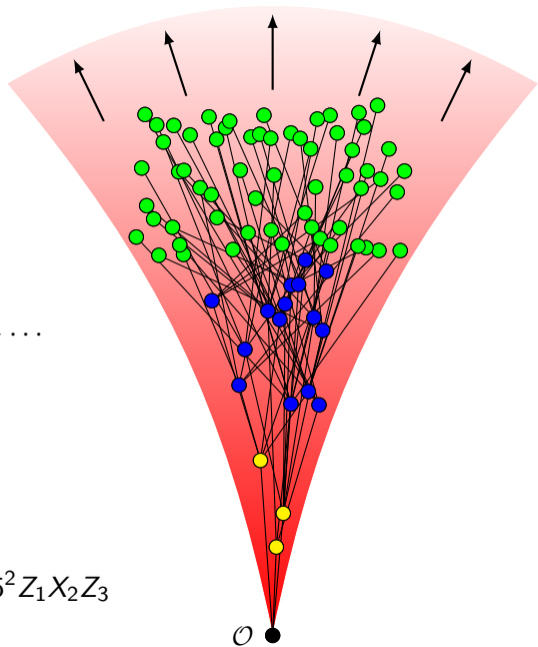
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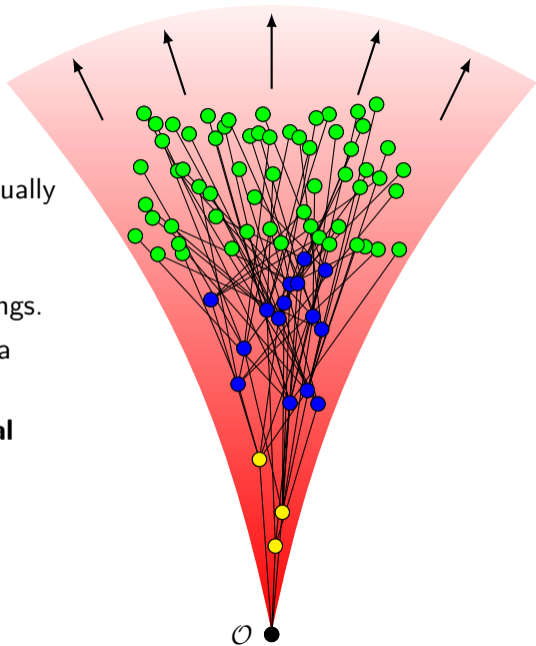
$$+ 1.05^2 Z_0 X_1 Z_2 + 1.05^2 X_1 + 1.05^2 X_2 + 1.05^2 Z_1 X_2 Z_3$$

$$+ 0.525 X_1 Z_2 + 0.525 Z_1 X_2 + 0.25 X_1.$$



The Basic Idea

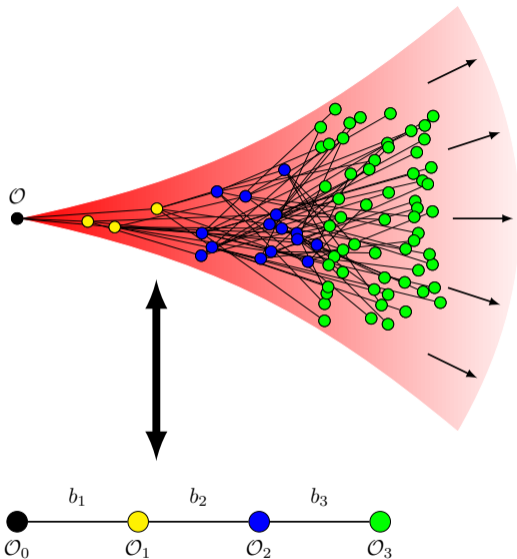
- ▶ Operators flow from simple to complex, eventually becoming too complex to compute.
- ▶ Complex operators are superpositions of a *thermodynamically large* number of Pauli strings.
- ▶ A sufficiently complex operator should admit a *universal* description.
- ▶ **Our goal now is to formulate this universal description.**



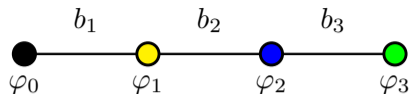
The Lanczos Algorithm

- ▶ Take the sequence $\{\mathcal{O}, \mathcal{L}\mathcal{O}, \mathcal{L}^2\mathcal{O}, \dots\}$ and apply Gram-Schmidt to orthogonalize $\{\mathcal{O}_0, \mathcal{O}_1, \mathcal{O}_2, \dots\}$.
- ▶ Explicitly, $|\mathcal{O}_1\rangle := b_1^{-1} \mathcal{L} |\mathcal{O}_0\rangle$, $b_1 := (\mathcal{O}_0 \mathcal{L} | \mathcal{L} \mathcal{O}_0)^{1/2}$,
 $|A_n\rangle := \mathcal{L} |\mathcal{O}_{n-1}\rangle - b_{n-1} |\mathcal{O}_{n-2}\rangle$,
 $b_n := (A_n | A_n)^{1/2}$ **“Lanczos Coefficients”**
 $|\mathcal{O}_n\rangle := b_n^{-1} |A_n\rangle$ **“Krylov vectors”**
- ▶ The Liouvillian is tridiagonal in this basis

$$L_{nm} := (\mathcal{O}_n^\dagger | \mathcal{L} | \mathcal{O}_m) = \begin{pmatrix} 0 & b_1 & 0 & 0 & \dots \\ b_1 & 0 & b_2 & 0 & \dots \\ 0 & b_2 & 0 & b_3 & \dots \\ 0 & 0 & b_3 & 0 & \dots \\ \vdots & \vdots & \vdots & \ddots & \ddots \end{pmatrix}$$



The Recursion Method



Define the **1D wavefunction** by

$$|\mathcal{O}(t)\rangle = \sum_{n=0}^{\infty} \varphi_n(t) |\mathcal{O}_n\rangle, \quad \varphi_n(t) := (\mathcal{O}_n | \mathcal{O}(t)).$$

The operator evolves as $-i\frac{d}{dt}\mathcal{O} = \mathcal{L}\mathcal{O}$, and \mathcal{L} is tridiagonal:

$$-i\partial_t \varphi_n = b_{n+1}\varphi_{n+1} + b_n\varphi_{n-1}, \quad \varphi_n(0) = \delta_{n0}.$$

The autocorrelation is just the wavefunction on site zero:

$$C(t) = (\mathcal{O}_0 | \mathcal{O}(t)) = \varphi_0(t).$$

This is called the **recursion method** and dates back to the 1980s.

Encodings of Dynamics

A. Correlation Function

$$C(t) := (\mathcal{O} | e^{i\mathcal{L}t} | \mathcal{O}) = \sum_{n=0}^{\infty} \frac{\mu_{2n}}{(2n)!} (it)^{2n}$$

B. Green's Function

$$G(z) := (\mathcal{O} | \frac{1}{z - \mathcal{L}} | \mathcal{O}) = \sum_{n=0}^{\infty} \frac{\mu_{2n}}{z^{2n+1}}$$

C. Spectral Function

$$\Phi(\omega) := \sum_{E, E'} |\langle E | \mathcal{O} | E' \rangle|^2 \delta(\omega - (E - E'))$$

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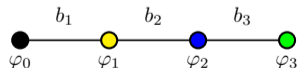
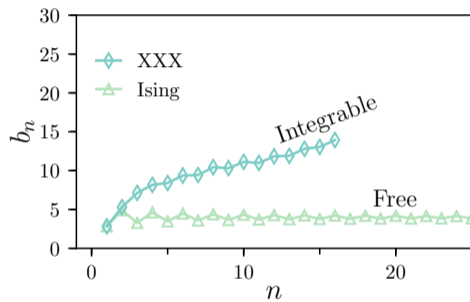
$$\Phi(\omega) := \sum_{E, E'} |\langle E | \mathcal{O} | E' \rangle|^2 \delta(\omega - (E - E'))$$

D. Lanczos Coefficients

$$\{b_n\}_{n=1}^{\infty} \quad \& \quad -i\partial_t \varphi_n = b_{n+1} \varphi_{n+1} + b_n \varphi_{n-1}$$

Empirical Patterns of Dynamics

Asymptotic	Growth Rate	System Type
$b_n \sim O(1)$	constant	Free models
$b_n \sim O(\sqrt{n})$	square-root	Integrable models
$b_n \sim ???$???	Chaotic models
$b_n \not\sim O(n)$	superlinear	Disallowed



Chaotic Examples

$$H_1 = \sum_i X_i X_{i+1} + 0.709 Z_i + 0.9045 X_i$$

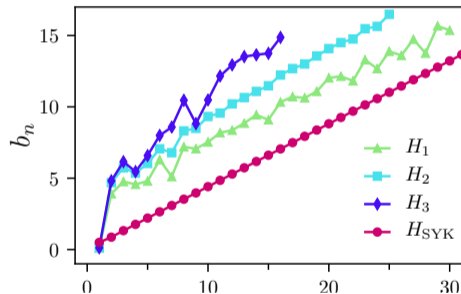
$$H_2 = H_1 + \sum_i 0.2 Y_i$$

$$H_3 = H_1 + \sum_i 0.2 Z_i Z_{i+1}$$

$$H_{\text{SYK}}^{(q)} = j^{q/2} \sum_{1 \leq i_1 < i_2 < \dots < i_q \leq N} J_{i_1 \dots i_q} \gamma_{i_1} \dots \gamma_{i_q},$$

$$\overline{J_{i_1 \dots i_q}^2} = 0,$$

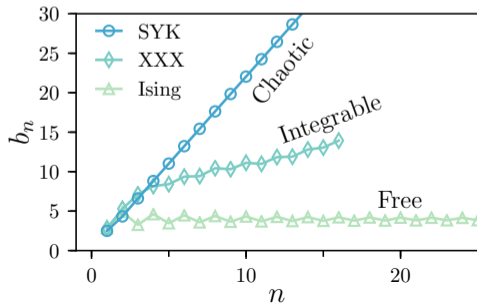
$$\overline{J_{i_1 \dots i_q}^2}^2 = \frac{(q-1)!}{N^{q-1}} J^2$$



Hypothesis: In a chaotic¹ quantum system, the Lanczos coefficients b_n are asymptotically linear, i.e. for $\alpha, \gamma \geq 0$,

$$b_n \xrightarrow{n \gg 1} \alpha n + \gamma.$$

Asymptotic	Growth Rate	System Type
$b_n \sim O(1)$	Constant	Free models
$b_n \sim O(\sqrt{n})$	Square-root	Integrable models
$b_n \sim O(n)$	Linear	Chaotic models
$b_n \not\sim O(n)$	Superlinear	Disallowed

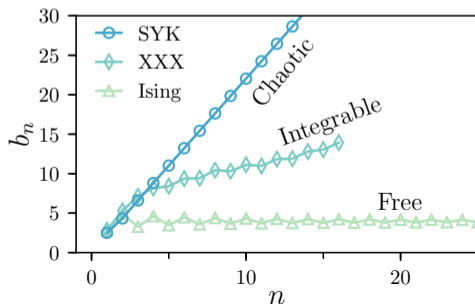


¹i.e. non-integrable

Hypothesis: In a chaotic quantum system, the Lanczos coefficients have asymptotics

$$b_n = \begin{cases} A \frac{n}{W(n)} + O(1) \sim \frac{An}{\log n} & \text{if } d = 1 \\ \alpha n + \gamma + o(1) \sim \alpha n & \text{if } d \geq 2 \end{cases}$$

Asymptotic	Growth Rate	System Type
$b_n \sim O(1)$	Constant	Free models
$b_n \sim O(\sqrt{n})$	Square-root	Integrable models
$b_n \sim O(n)$	Linear	Chaotic models
$b_n \not\sim O(n)$	Superlinear	Disallowed



Thermalization

The Hypothesis enforces “irreversible” dynamics.

Exact Asymptotic Behavior

Model

$$\tilde{b}_n := \alpha \sqrt{n(n-1+\eta)} \xrightarrow{n \gg 1} \alpha n + \gamma.$$

Exact solution

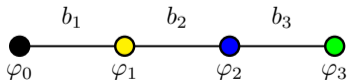
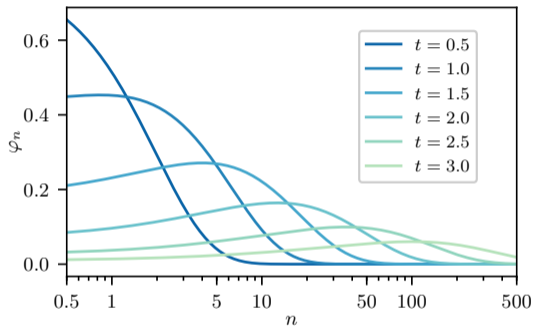
$$\tilde{\varphi}_n(t) = \sqrt{\frac{(\eta)_n}{n!}} \tanh(\alpha t)^n \operatorname{sech}(\alpha t)^\eta$$

where $(\eta)_n = \eta(\eta+1)\cdots(\eta+n-1)$.

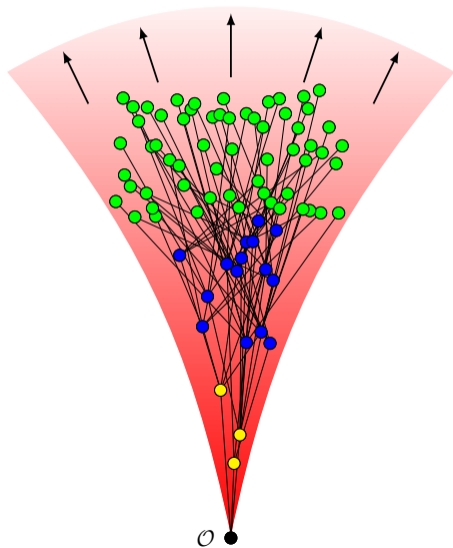
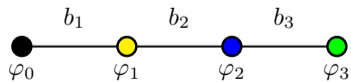
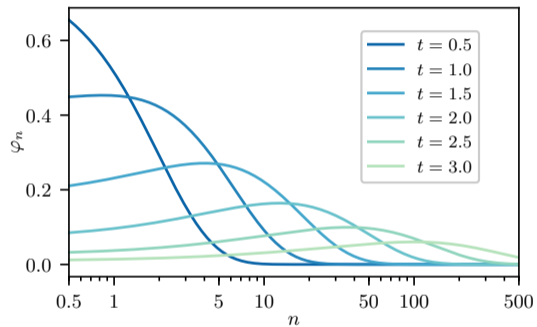
Define the **Krylov space position** operator

$$(n(t)) := \sum_{n \in \mathbb{N}} n |\varphi_n(t)|^2 = \eta \sinh(\alpha t)^2 \sim e^{2\alpha t}$$

The wavefunction runs away “irreversibly” into the 1D chain.

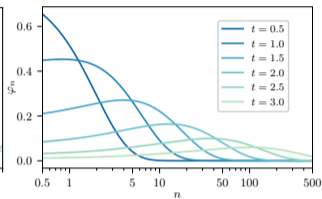
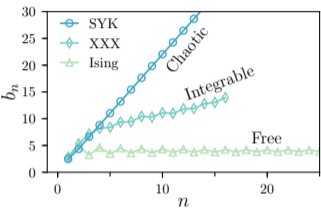


Picture of Thermalization



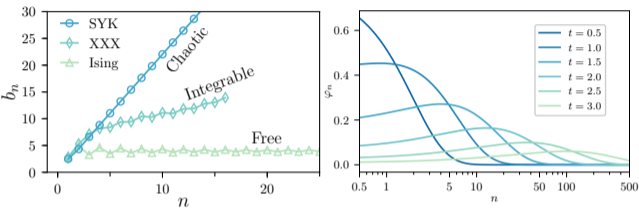
Four Observables

D. Lanczos Coefficients $b_n \sim \alpha n$

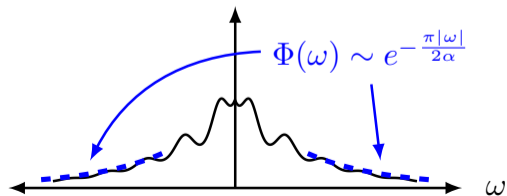


Four Observables

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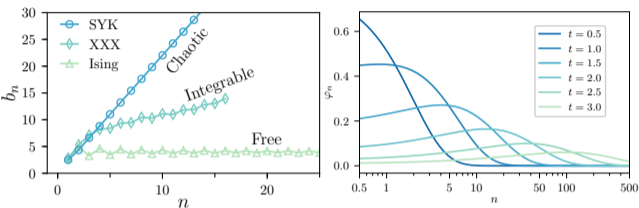


C. Spectral Function



Four Observables

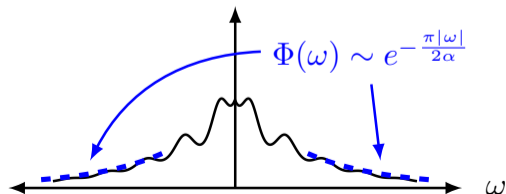
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B. Green's Function

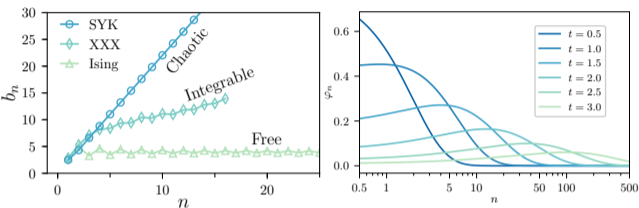
$$G(z) = \frac{1}{z - \frac{b_1^2}{z - \frac{b_2^2}{z - \frac{b_3^2}{\ddots}}}}$$

C. Spectral Function

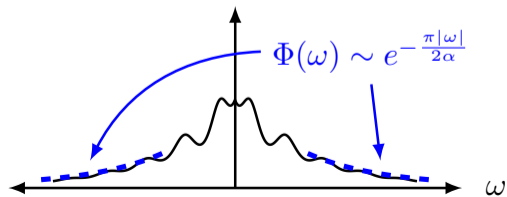


Four Observables

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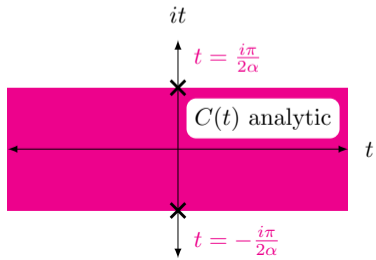
C. Spectral Function



B. Green's Function

$$G(z) = \frac{1}{z - \frac{b_1^2}{z - \frac{b_2^2}{z - \frac{b_3^2}{\dots}}}}$$

A. Correlation Function



Quantum Chaos

The Lanczos coefficients “diagnose” quantum chaos.

Hallmarks of Quantum Chaos

- ▶ **Level Statistics** give a “highly microscopic” indicator of chaos. With $\delta E_n := E_{n+1} - E_n$, the r-statistic is:

$$r = \frac{\overline{\min \{\delta E_n, \delta E_{n+1}\}}}{\overline{\max \{\delta E_n, \delta E_{n+1}\}}} \approx \begin{cases} 0.386 & \text{integrable} \\ 0.530 & \text{chaotic} \end{cases}$$

- ▶ **Eigenstate Thermalization Hypothesis** predicts matrix elements of chaotic systems

$$\langle E_n | \mathcal{O} | E_m \rangle = \mathcal{O}(\bar{E}) \delta_{nm} + e^{-S(\bar{E})} f_{\mathcal{O}}(\bar{E}, \omega) R_{nm}$$

Dynamics	Level Stats.	b_n	$\Phi(\omega)$ or $ f_{\mathcal{O}} ^2$
Free	-	$O(1)$	$\theta(\omega - 2W)$
Integrable	Poisson	$O(\sqrt{n})$	$O(e^{-\omega^2})$
Chaotic	GOE	$O(n)$	$O(e^{-\omega})$

- ▶ **Exponential sensitivity**

Example 1: Chaotic Ising Model

Model

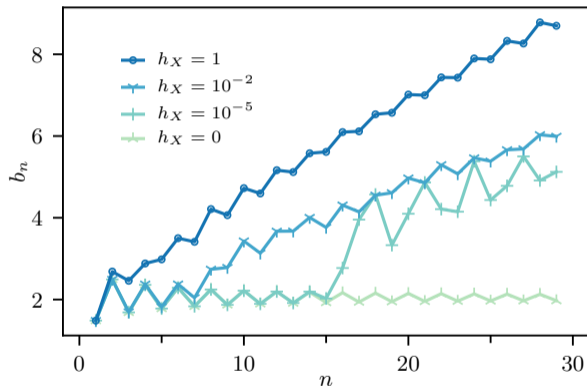
$$H = \sum_i J [X_i X_{i+1} + h Z_i] + h_x X_i$$

Dynamics In the thermodynamic limit,

$$\begin{cases} h_x = 0 & \text{free model} \\ h_x > 0 & \text{chaotic model.} \end{cases}$$

Perturbation Theory Resonances appear at order $O(h_x/J)$.

$$b_n \approx \begin{cases} O(1) & n < O(h_x/J) \\ O(n) & n > O(h_x/J) \end{cases}$$



Example 2: SYK Model

Model¹

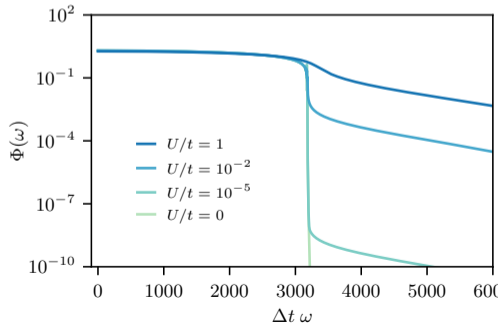
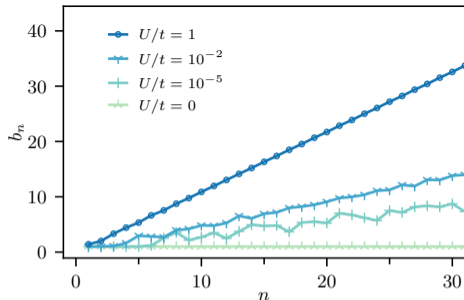
$$H = tH_{\text{SYK}}^{(2)} + UH_{\text{SYK}}^{(4)}$$

$$H_{\text{SYK}}^{(q)} = i^{q/2} \sum_{1 \leq i_1 < i_2 < \dots < i_q \leq N} J_{i_1 \dots i_q} \gamma_{i_1} \dots \gamma_{i_q},$$

$$\overline{J_{i_1 \dots i_q}^2} = 0, \quad \overline{J_{i_1 \dots i_q}^2}^2 = \frac{(q-1)!}{N^{q-1}} j^2$$

Dynamics

$$q = \begin{cases} 2 & \text{free model} \\ 4, 6, 8, \dots & \text{chaotic model} \end{cases}$$



¹Sachdev, Ye, 1993; Parcollet, Georges, 1999; Kitaev, 2015; Maldacena, Stanford, 2016, etc.

Example 3: XXZ+NNN

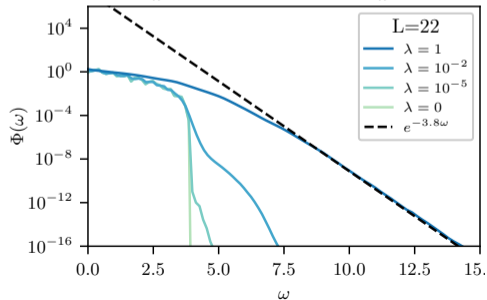
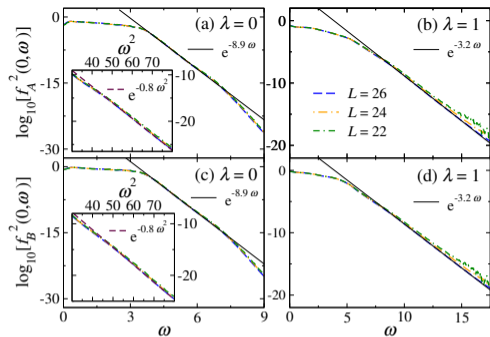
Model

$$H_1 = \sum_i X_i X_{i+1} + Y_i Y_{i+1} + \Delta Z_i Z_{i+1} \\ + \lambda \left(\sum_i X_i X_{i+2} + Y_i Y_{i+2} + \Delta Z_i Z_{i+2} \right)$$

$$A := \sum_i Z_i Z_{i+1}, B := \sum_i S_i^+ S_{i+2}^- + \text{h.c.}$$

Dynamics

$$\begin{cases} \Delta = 0 \ \& \ \lambda = 0 & \text{free model} \\ \Delta \neq 0 \ \& \ \lambda = 0 & \text{integrable model} \\ \Delta \neq 0 \ \& \ \lambda \neq 0 & \text{chaotic model} \end{cases}$$



Model	Op.	Dynamics	Lanczos	Evidence	Ref.
Ising	\widehat{Z}	Free	$O(1)$	Analytic	Viswanath & Müller
XX	\widehat{Z}	Free	$O(1)$	Analytic	Viswanath & Müller
SYK ⁽²⁾	γ	Free	$O(1)$	Analytic	Maldacena, Shenkar, Stanford, 2016
XX	\widehat{X}	Free*	$O(\sqrt{n})$	Analytic	Viswanath & Müller
Free Fermions in Disguise	\widehat{Z}	Free*	$O(\sqrt{n})$	Numerical	see Fendley, 2019.
MBL	\widehat{Z}	Int.	$O(\sqrt{n})$	Numerical	
XXZ	\widehat{Z}	Int.	$O(\sqrt{n})$	Numerical	
Chaotic Ising	\widehat{Z}	Chaotic	$O(n)$	Numerical	
XXZ + NNN	\widehat{ZZ}	Chaotic	$O(n)$	Numerical	LeBlond, Mallayya, Vidmar, Rigol, 2019.
SYK ⁽⁴⁾	γ	Chaotic	$O(n)$	Numerical	
SYK ^(∞)	γ	Chaotic	$O(n)$	Analytic	Roberts, Stanford, Streicher, 2018.
SYK Hopping	γ	Chaotic	$O(n)$	Analytic	
2D Fermi Hubbard	\widehat{J}	Chaotic	$O(n)$	Numerical	Huang, private comm.
Bouch Model	\widehat{X}	Chaotic	$O(n)$	Analytic	Bouch, 2015

Complexity

The Lanczos coefficients quantify quantum chaos.

K-Complexity

I will introduce a measure of quantum chaos (“**K-Complexity**”) that is

1. easy to interpret
2. easy to compute
3. works in all quantum systems (not semiclassical).

Exponential Sensitivity

- ▶ A hallmark of chaos is *exponential sensitivity* to small perturbations.
- ▶ Classically, this is measured by the Lyapunov exponent.

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- ▶ The **Out-of-time-order commutator** generalizes the Lyapunov exponent λ_L to semi-classical systems¹

$$\text{OTOC}(t) := ([\mathcal{O}(t), V][\mathcal{O}(t), V]) \sim e^{\lambda_L t}.$$

and λ_L is called the **quantum Lyapunov exponent**.

¹ Kitaev, 2014. ² Maldacena, Shenker, Stanford, 2016.

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and λ_L is called the **quantum Lyapunov exponent**.

- ▶ High-energy theorists have shown a “universal bound on chaos”: for $T \rightarrow 0$,²

$$\lambda_L \leq 2\pi T. \tag{1}$$

¹ Kitaev, 2014. ² Maldacena, Shenker, Stanford, 2016.

Out-of-time-order Confusion

Semiclassical OTOCs usually saturate at¹

$$t = \begin{cases} O(\log(1/\hbar)) & \text{semiclassics} \\ O(\log N) & \text{large-}N \\ O(1) & \text{no small parameter} \end{cases}$$

Regularization Dependent Choice of norm:

$$(A|B)_\beta := \text{Tr}[\rho A^\dagger B] \quad \text{“Physical”}$$

$$(A|B)_\beta^W := \text{Tr}[\rho^{1/2} A^\dagger \rho^{1/2} B] \quad \text{“Wightman”}$$

where $\rho = e^{-\beta H} / Z$.

¹ Maldacena, Shenker, Stanford, 2018. ² Khemani, Huse, Nahum, 2018.

Out-of-time-order Confusion

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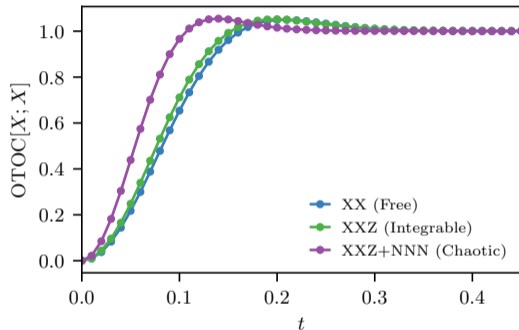
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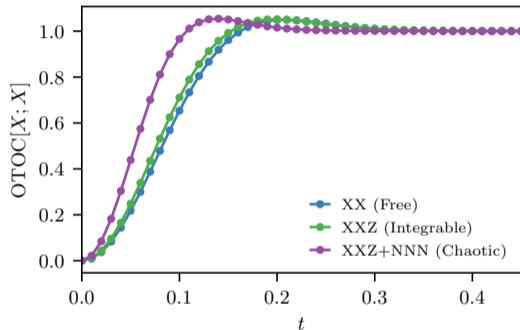
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where $\rho = e^{-\beta H}/Z$.



Therefore:

- ▶ Difficult to define λ_L in spin chains.
- ▶ Must use clever tricks like velocity-dependent Lyapunov exponents.²

¹ Maldacena, Shenker, Stanford, 2018. ² Khemani, Huse, Nahum, 2018.

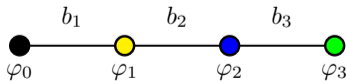
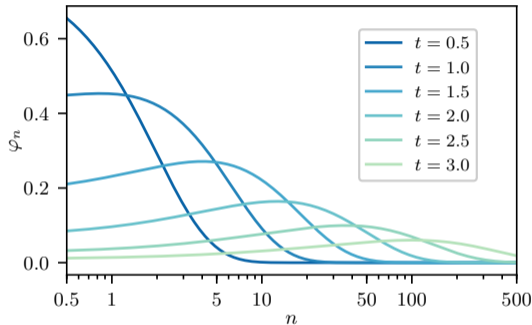
K-Complexity

The Krylov vectors \mathcal{O}_n grow successively larger, have more components, need more resources. . . they are more complex.

Therefore define the **K-Complexity** as

$$(n(t)) := \sum_{n \in \mathbb{N}} n |\varphi_n(t)|^2 \sim e^{2\alpha t}$$

where $\varphi_n(t) := (\mathcal{O}_n | \mathcal{O}(t))$.



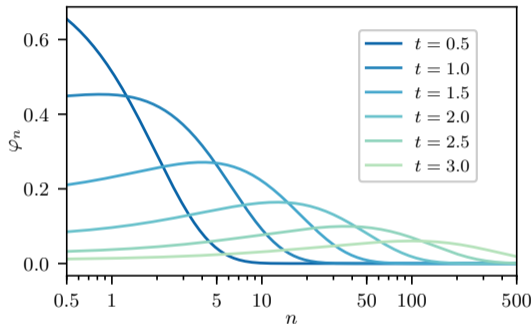
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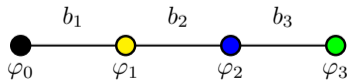
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α/\mathcal{J}	0	0.461	0.623	0.800	0.863	1
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¹Roberts, Stanford, Streicher, 2018.

Rigorous Bounds

Proposition: Suppose $T = \infty$. For any local operator, $\exists C > 0$ such that

$$\text{OTOC}(t) \leq C \cdot (n(t)).$$

Rigorous Bounds

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$$\text{OTOC}(t) \leq C \cdot (n(t)).$$

Corollary: Suppose $b_n \asymp \alpha n$ and $T = \infty$. Then, if the quantum Lyapunov exponent λ_L is defined,

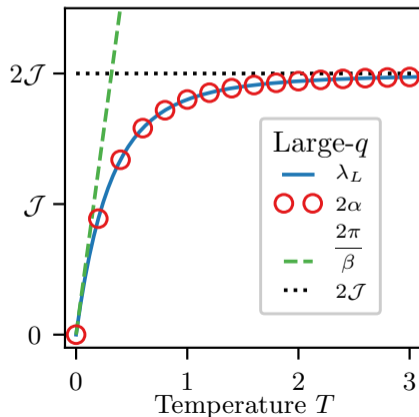
$$\lambda_L \leq 2\alpha.$$

Proposition: Suppose $b_n \asymp \alpha_W n$ using the Wightman regularization. Then

$$2\alpha \leq 2\pi T.$$

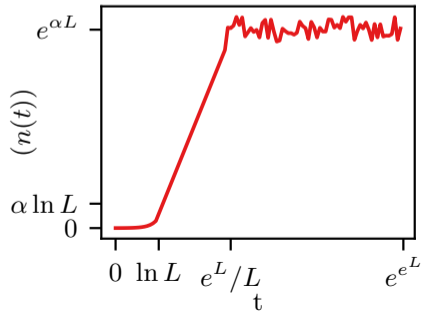
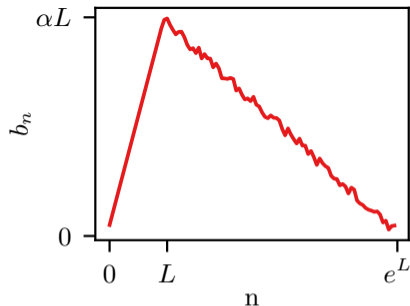
Conjecture: Under the same assumptions,

$$\lambda_L \leq 2\alpha \leq 2\pi T.$$



Finite Systems

- ▶ The Krylov operators \mathcal{O}_n hit the edge at $n = O(L)$ and stop growing.
- ▶ Random matrix theory then kicks in and orthogonality causes the b_n 's to decrease.
- ▶ The K -complexity keeps growing linearly until $t = O(e^L/L)$.
- ▶ This seems to match quantitatively with the “switchback effect” considered in black hole complexity.^{1,2}



¹Barbón, Rabinovici, Shir, Sinha, 2019. ²DP *et al*, in progress.

Chaos and Complexity

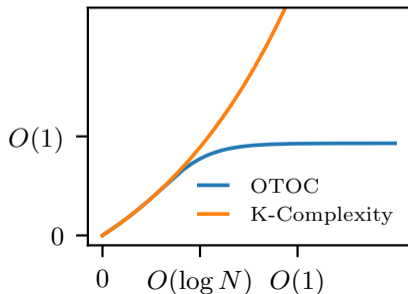
$$(n(t)) := \sum_{n \in \mathbb{N}} n |\varphi_n(t)|^2 \asymp e^{2\alpha t}$$

Dynamics	Level Stats.	b_n	$\Phi(\omega)$ or $ f_0 ^2$	K-complexity
Free	-	$O(1)$	$\theta(\omega - 2W)$	$(n)_t \asymp t$
Integrable	Poisson	$O(\sqrt{n})$	$O(e^{-\omega^2})$	$(n)_t \asymp t^2$
Chaotic	GOE	$O(n)$	$O(e^{-\pi\omega/2\alpha})$	$(n)_t \asymp e^{2\alpha t}$

Therefore α is the [complexity growth rate](#).

This is [experimentally observable](#) from high-frequency heating.

The complexity is [non-perturbative data](#) needed to compute hydrodynamics.



Quantum Mechanics



Operator
Growth
Hypothesis



Thermalization



Chaos



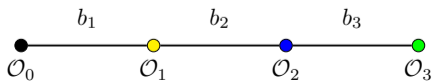
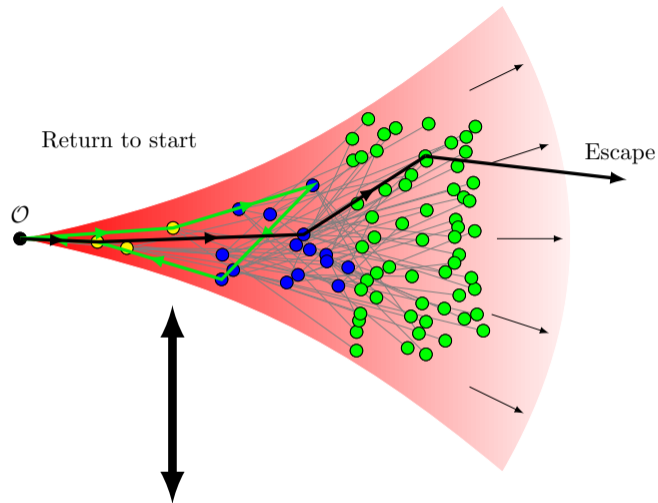
Complexity

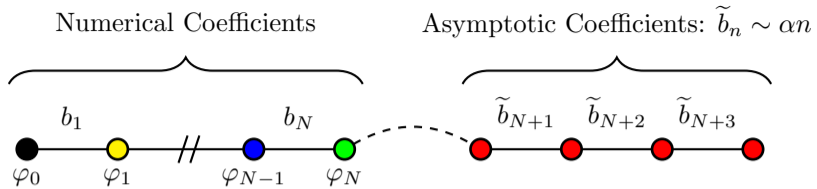


Hydrodynamics

Hydrodynamics

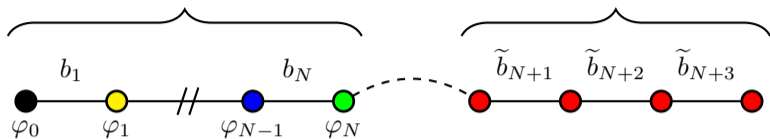
K-complexity gives rise to emergent hydrodynamics.





Numerical Coefficients

Asymptotic Coefficients: $\tilde{b}_n \sim \alpha n$



$$L = \begin{pmatrix} 0 & b_1 & 0 & 0 \\ b_1 & 0 & b_2 & 0 \\ 0 & b_2 & 0 & \ddots \\ 0 & 0 & \ddots & \ddots \end{pmatrix}$$

$$\approx \begin{pmatrix} 0 & b_1 & 0 & 0 \\ b_1 & 0 & \ddots & 0 \\ 0 & \ddots & \ddots & b_N \\ 0 & 0 & b_n & \widetilde{G^{(N)}}(z) \end{pmatrix}$$

$$G(z) = \int dt e^{izt} \langle \mathcal{O}(t) \mathcal{O}(0) \rangle$$

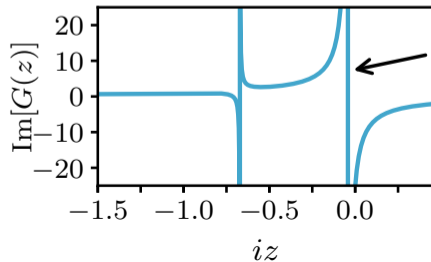
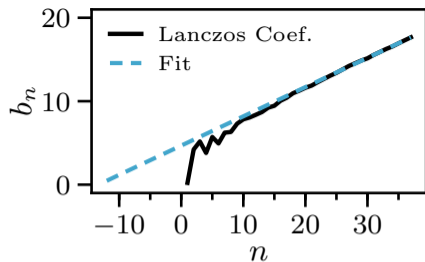
$$\approx \frac{1}{z - \frac{b_1^2}{z - \frac{b_2^2}{z - \frac{b_2^2}{\ddots}}}}}$$

$$\widetilde{G^{(N)}}(z) = \Gamma(N+1) \Gamma\left(\frac{z+1}{2}\right)$$

$$\times {}_2F_1\left(N+1, \frac{z+1}{2}, \frac{z+2N+3}{2}; -1\right)$$

Algorithm

0. Choose a local operator \mathcal{O} whose correlation $C(t) = \text{Tr}[\mathcal{O}(t)\mathcal{O}(0)]$ should be hydrodynamical.
1. Compute b_1, \dots, b_N via infinite exact diagonalization and fit the slope α .
2. Stitch together the b_n 's and the asymptotic solution $\widetilde{G}^{(N)}$.
3. Identify the pole closest to the origin to extract the hydrodynamical dispersion relation.



Diffusion in the Chaotic Ising Model

Chaotic Ising Model

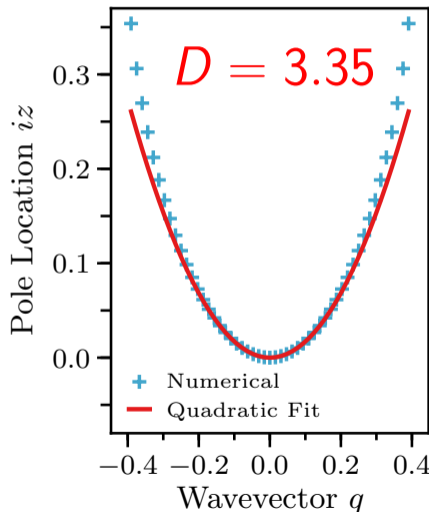
$$H = \sum_j X_j + 1.05Z_jZ_{j+1} + 0.5Z_j$$

Initial operator at wavevector k :

$$\mathcal{O}_k = \sum_j e^{ikj} (X_j + 1.05Z_jZ_{j+1} + 0.5Z_j)$$

We see the dispersion relation for diffusion

$$\frac{d}{dt}\epsilon(t, x) = D\nabla^2\epsilon(t, x).$$



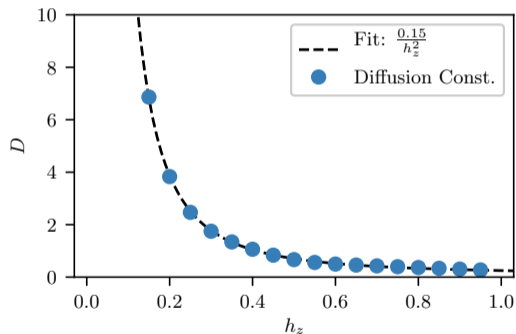
Diffusive Phenomenology

Chaotic Ising Model

$$H = \sum_j X_j + 1.05 Z_j Z_{j+1} + h_x Z_j$$

As $h_x \rightarrow 0$, a Drude peak ($D \rightarrow \infty$) emerges.

This is a practical method for computing hydrodynamics at strong coupling.



Finite Temperature

At $T < \infty$ there is a *physical* choice of inner product.

Suppose g is an even measure on the thermal circle:

1. $g : [0, \beta] \rightarrow \mathbb{R}$ (or a distribution)
2. $g(\lambda) = g(\beta - \lambda)$
3. $\int_0^\beta g(\lambda) = 1$.

Then, with $y := e^{-H}$, there is a g -inner product:

$$(A|B)_\beta^g := \frac{1}{Z(\beta)} \int_0^\beta g(\lambda) \operatorname{Tr}[y^{\beta-\lambda} A^\dagger y^\lambda B] d\lambda \quad (3)$$

Two common choices are:

Physical $(A|B)_\beta^P := Z^{-1} \operatorname{Tr}[\rho A^\dagger B] + (A \leftrightarrow B)$

Wightman $(A|B)_\beta^W := Z^{-1} \operatorname{Tr}[\rho^{1/2} A^\dagger \rho^{1/2} B] + (A \leftrightarrow B)$

In the limit $T \rightarrow \infty$ or $\beta \rightarrow 0$, these all give the Hilbert-Schmidt inner product.

Quantum Mechanics



Operator
Growth
Hypothesis



Thermalization



Hydrodynamics



Complexity



Chaos

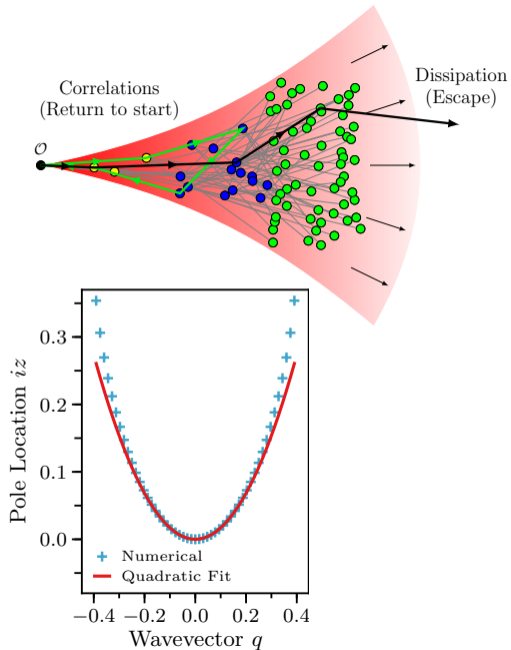
Summary: [arXiv: 1812.08657](https://arxiv.org/abs/1812.08657)

- ▶ The hypothesis governs operator growth in chaotic, closed quantum systems

$$b_n = \begin{cases} A \frac{n}{W(n)} + O(1) \sim \frac{An}{\log n} & \text{if } d = 1 \\ \alpha n + \gamma + o(1) \sim \alpha n & \text{if } d \geq 2 \end{cases}$$

- ▶ Emergence of hydrodynamics in a computationally tractable scheme.
- ▶ The **operator growth rate** α also controls the growth of complexity and chaos in quantum systems: $\lambda_L \leq 2\alpha$

SYK- q	2	3	4	7	10	∞
$2\alpha/\mathcal{J}$	0	0.461	0.623	0.800	0.863	1
$\lambda_L/(2\mathcal{J})$	0	0.454	0.620	0.799	0.863	1



Future Work

- ▶ Log corrections disrupt asymptotics in 1D. How does our numerical technique still work in 1D?
- ▶ Can we prove the hypothesis within random matrix theory?
- ▶ How can we extend the hypothesis to finite temperature?
- ▶ Can we compute b_n in QMC or other numerical techniques in 2D?
- ▶ Can we say anything about the MBL transition with this notion of chaos/ergodicity?
- ▶ Can we measure α experimentally? Perhaps from $\Phi(\omega)$ at large ω ?

Extra Slides

Translation-Invariant MBL (Preliminary)

Ising MBL model:

$$H = \sum_j Z_j Z_{j+1} + 1.05 X_j + W_j Z_j.$$

To recover translation-invariance, promote

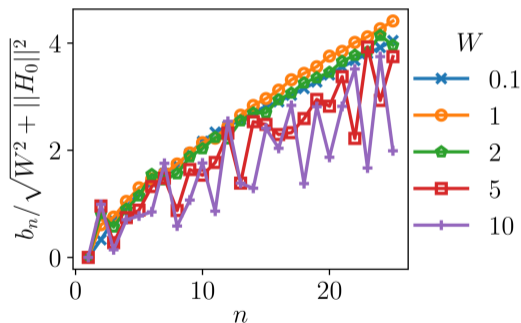
$$W_j \in \{-1, 1\} \rightarrow W \hat{\tau}_j^z$$

where $\hat{\tau}_j^z$ is a “binary disorder operator”.

$$\tilde{H} = \sum_j Z_j Z_{j+1} + 1.05 X_j + W \hat{\tau}_j^z Z_j.$$

Pro: can compute Lanczos *directly in the thermodynamic limit*.

Con: doubled on-site operator space.



Translation-Invariant ℓ -bits (Preliminary)

Start with $\mathcal{O} = Z$.

1. Truncate in Krylov space:

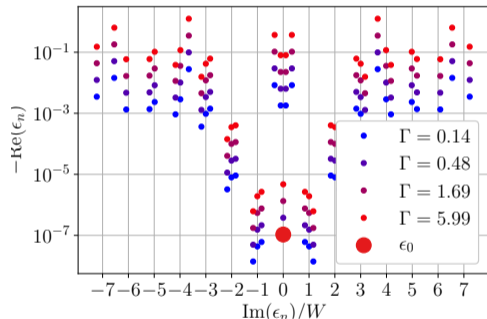
$$T(\Gamma) := \begin{pmatrix} 0 & b_1 & 0 & 0 & \cdots \\ b_1 & 0 & b_2 & 0 & \cdots \\ 0 & b_2 & 0 & \cdots & \cdots \\ 0 & 0 & \cdots & 0 & b_N \\ \vdots & \vdots & \vdots & b_N & -i\Gamma \end{pmatrix}.$$

2. Solve

$$T\mathcal{A}_\alpha = \varepsilon_\alpha \mathcal{A}_\alpha$$

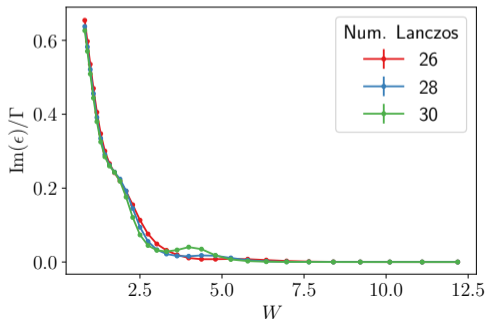
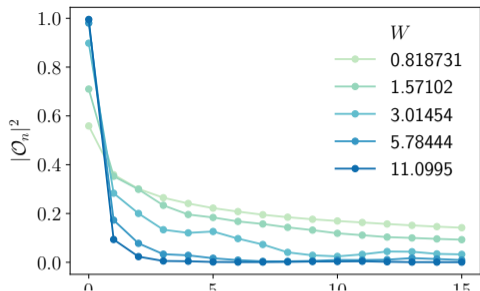
3. Smallest eigenvalue $\varepsilon_0(\Gamma) = 0 - iE_0\Gamma$
with \mathcal{A}_0 well-localized to small n .

Interpretation: \mathcal{A}_0 is a “translation- and disorder-averaged ℓ -bit”.



MBL Transition (Preliminary)

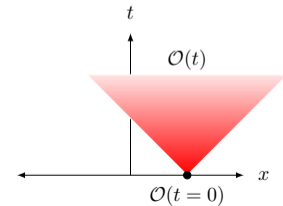
- ▶ Can we see the MBL transition?
- ▶ For large- W , the “ l -bit” is exponentially localized in the chain.
- ▶ For small- W , it decays as a power-law (fairly generic in ETH systems).
- ▶ **Interpretation:** Lanczos gives a non-perturbative probe of the MBL transition.



History

Mathematical Results

- ▶ Araki (1969)
- ▶ Lieb-Robinson Bound (1972)
- ▶ ETH (1994)
- ▶ ADHH Theorem (2015)



OTOCs

- ▶ Quantum version of Lyapunov exponent (Kitaev)
- ▶ Maldacena-Shenkar-Stanford bound at low- T (2015)
- ▶ Computable in SYK, large- N , holography. . .
- ▶ Only well-defined semiclassically

Random Unitaries

- ▶ Solvable models of quantum chaos
- ▶ Local, finite- N , operator front propagation
- ▶ Emergent dissipation
- ▶ Non-Hamiltonian dynamics, no Lyapunov exponents
- ▶ (Nahum, Khemani, Huse, Pollmann, etc)

The Lanczos Algorithm

The Lanczos algorithm iteratively *tridiagonalizes* a matrix

Algorithm:

1. Define

$$|\mathcal{O}_0\rangle := \mathcal{O}, b_0 := 0$$

2. For each n , apply \mathcal{L} to make a new operator:

$$|A_n\rangle := \mathcal{L}|\mathcal{O}_{n-1}\rangle - b_{n-1}|\mathcal{O}_{n-2}\rangle$$

3. Orthogonalize again previous operator:

$$|\mathcal{O}_n\rangle := b_n^{-1}|A_n\rangle, b_n := (A_n|A_n)^{1/2}$$

4. Repeat until $|\mathcal{O}_n\rangle$ vanishes.

The Liouvillian becomes **tridiagonal**

$$L_{nm} := (\mathcal{O}_n|\mathcal{L}|\mathcal{O}_m) = \begin{pmatrix} 0 & b_1 & 0 & 0 & \dots \\ b_1 & 0 & b_2 & 0 & \dots \\ 0 & b_2 & 0 & b_3 & \dots \\ 0 & 0 & b_3 & 0 & \ddots \\ \vdots & \vdots & \vdots & \ddots & \ddots \end{pmatrix}$$

The b_n 's are called **Lanczos coefficients** and the $|\mathcal{O}_n\rangle$'s are called **Krylov vectors**.

Log Corrections in 1D

Theorem (Araki 1969) For any Hamiltonian with local interactions

$$C(t + i\tau) = \langle \mathcal{O} | e^{i\mathcal{L}(t+i\tau)} | \mathcal{O} \rangle$$

is an entire function of $t + i\tau \in \mathbb{C}$.

Corollary The asymptotic growth of the Lanczos coefficients is strictly sublinear in one dimension. In fact,

$$b_n \leq A \frac{n}{W(n)}$$

where W is the product-log function defined by $z = W(ze^z)$ whose asymptotic is $W(n) \sim \ln n - \ln \ln n + O(1)$.

Therefore the hypothesis is modified in 1D. We still permit $b_n \geq n^\alpha$ for any $\alpha < 1$.

Higher Dimensions

Theorem (Bouch 2011) For $d = 2$ (and higher), there exists a local Hamiltonian whose correlation function fail to be entire. Namely

$$H = \sum_{(x,y) \in \mathbb{Z}^2} Z_{x,y} X_{x+1,y} + X_{x,y} Z_{x,y-1}$$

with $\mathcal{O} = X_{0,0}$. (This achieves linear growth of b_n .)

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with $\mathcal{O} = X_{0,0}$. (This achieves linear growth of b_n .)

Corollary For $d \geq 2$, linear growth $b_n = \alpha n + O(1)$ is a tight upper bound for the growth of the Lanczos coefficients.

So the hypothesis survives unscathed in higher dimensions.

Eigenstate thermalization hypothesis:

$$\mathcal{O}_{\alpha\beta} = \mathcal{O}(\bar{E})\delta_{\alpha\beta} + e^{-S(\bar{E})}f_{\mathcal{O}}(\bar{E},\omega)R_{\alpha\beta} \quad (4)$$

where \mathcal{O} is a local observable, $\bar{E} = (E_{\alpha} + E_{\beta})/2$, $\omega = E_{\alpha} - E_{\beta}$, $S(\bar{E})$ is the thermodynamic entropy, $R_{\alpha\beta}$ a random variable and $\mathcal{O}(\bar{O})$ and $f_{\mathcal{O}}$ are smooth.

The operator growth hypothesis implies (at $T = \infty$)

$$\text{quantum chaos} \iff \int d\bar{E} f_{\mathcal{O}}(\bar{E},\omega) = e^{-\frac{\pi|\omega|}{2\alpha}} + \mathcal{O}(1).$$

Finite Temperature

At $T < \infty$ there is a *physical* choice of inner product.

Suppose g is an even measure on the thermal circle:

1. $g : [0, \beta] \rightarrow \mathbb{R}$ (or a distribution)
2. $g(\lambda) = g(\beta - \lambda)$
3. $\int_0^\beta g(\lambda) = 1$.

Then, with $y := e^{-H}$, there is a g -inner product:

$$(A|B)_\beta^g := \frac{1}{Z(\beta)} \int_0^\beta g(\lambda) \operatorname{Tr}[y^{\beta-\lambda} A^\dagger y^\lambda B] d\lambda \quad (5)$$

Two common choices are:

Physical $(A|B)_\beta^P := Z^{-1} \operatorname{Tr}[\rho A^\dagger B] + (A \leftrightarrow B)$

Wightman $(A|B)_\beta^W := Z^{-1} \operatorname{Tr}[\rho^{1/2} A^\dagger \rho^{1/2} B] + (A \leftrightarrow B)$

In the limit $T \rightarrow \infty$ or $\beta \rightarrow 0$, these all give the Hilbert-Schmidt inner product.