



RUTGERS
THE STATE UNIVERSITY
OF NEW JERSEY



Lecture 1: From Classical and Quantum Chaos to Thermalization in Isolated Quantum Systems

Jed Pixley

ICTS Lecture Series:

Quantum Dynamics in the pre-fault tolerant era

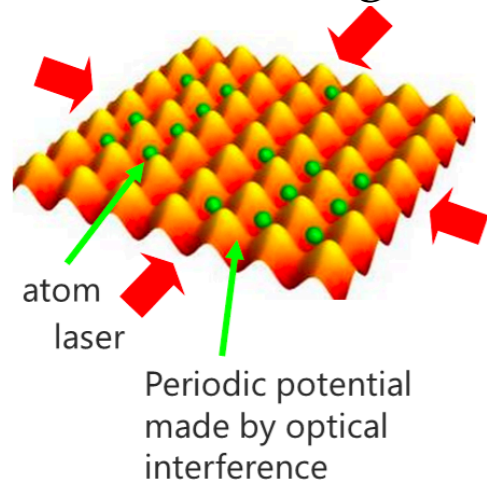
6/3/2026



QUANTUM TECHNOLOGY OF TODAY

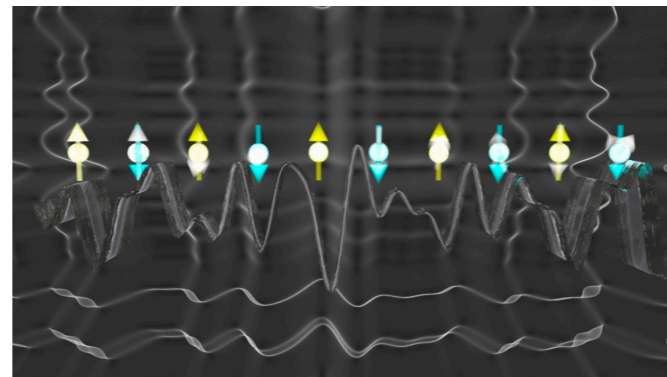
Huge advances in recent decades:

Cold atomic gases



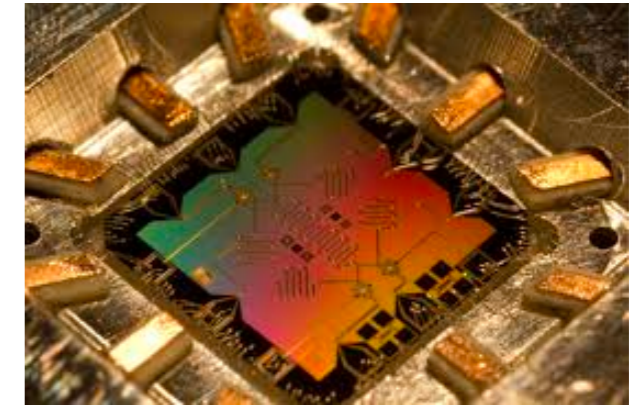
M. Kozuma

Trapped ions



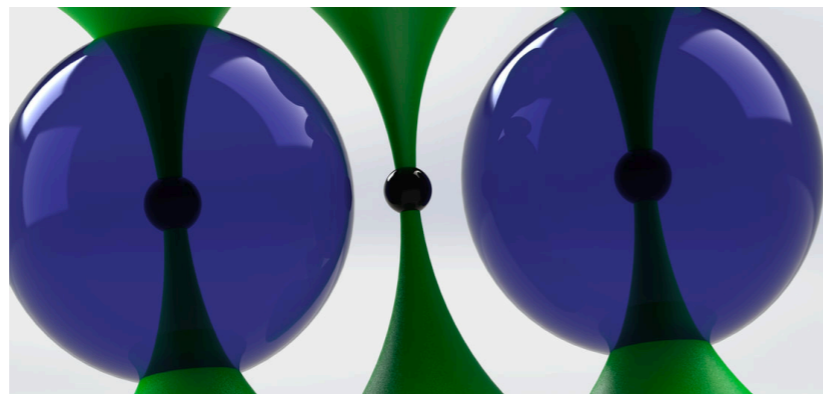
C. Monroe

Superconducting qubits



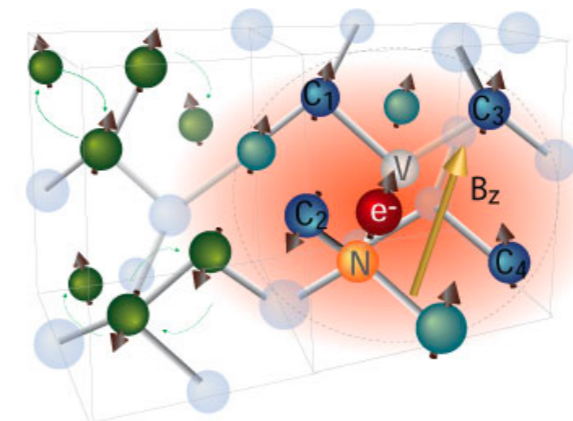
J. Martinis

Neutral atoms



M. Endres

Nitrogen-vacancy centers in diamond



P. Cappellaro

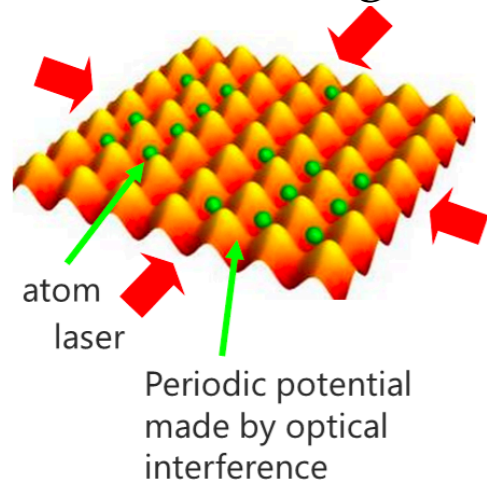
Highly coherent quantum systems with measurement capabilities

Applications: Computing, Simulation, Sensing...

QUANTUM TECHNOLOGY OF TODAY

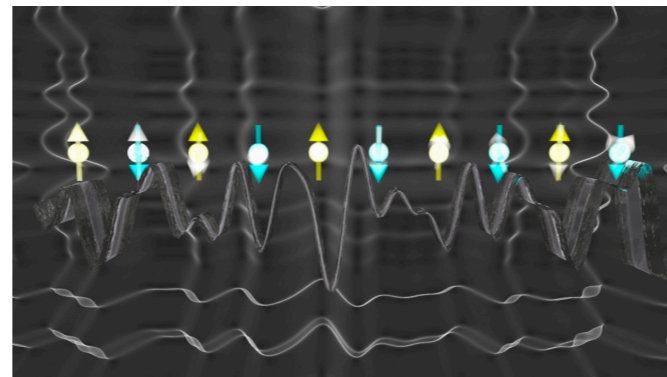
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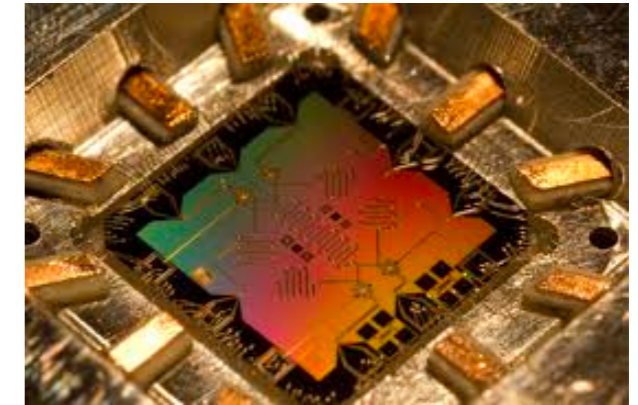
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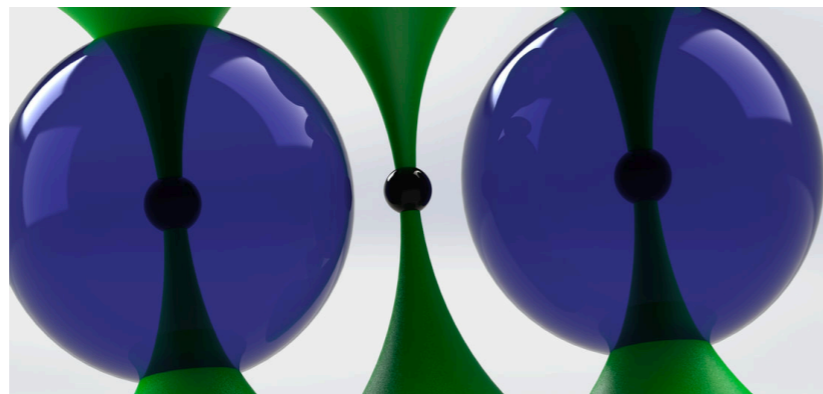
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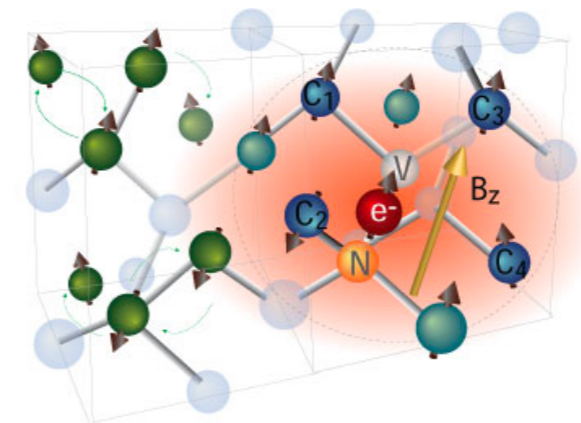
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P. Cappellaro

Generic quantum dynamics is chaotic/thermal, too hard to control in general.

How can we control and manipulate quantum systems?

LECTURE SERIES OVERVIEW

Quantum dynamics in the pre fault tolerant era

Lecture 1: From classical and quantum chaos to thermalization in isolated quantum systems

Lecture 2: Analog and digital quantum simulators from unitary dynamics to midcircuit measurements.

Lecture 3: Monitored quantum dynamics in random quantum circuits

Lecture 4: Adaptive quantum circuits and control induced phase transitions

Lecture 5: Open quantum dynamics software tutorial

LECTURE SERIES, LEARNING GOALS

- I. **Lecture 1: Classical and Quantum Chaos, from single particle to many-body**
- II. Lecture 2: Quantum platforms
- III. Lecture 3: Monitored dynamics, interplay of unitary and projective evolution.
- IV. Lecture 4: Adaptive dynamics, controlling quantum dynamics
- V. Lecture 5: Numerical approaches to adaptive quantum dynamics

LECTURE 1: GOALS

Understanding generic quantum dynamics

Classical dynamics, chaotic vs integrable

Quantum dynamics, chaos vs integrable

Thermalization in isolated quantum systems

Models for generators of the dynamics:

Hamiltonian, Floquet, and random quantum circuits

Breakdown of thermalization

OUTLINE

- I. Lecture series layout
- II. Classical chaos
- III. Quantum chaos
- IV. Many-body quantum chaos and thermalization
- V. Evading thermalization

UNPREDICTABILITY OF CHAOS

Classical Chaos: unpredictable deterministic evolution of the classical equations of motion.

Weather



Google

Public transit



Feldman, *Chaos and Fractals an Elementary Introduction*, (2012)

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3 gravitational bodies

Gravitational 3-body problem

Elastic 3-body problem



Wikipedia

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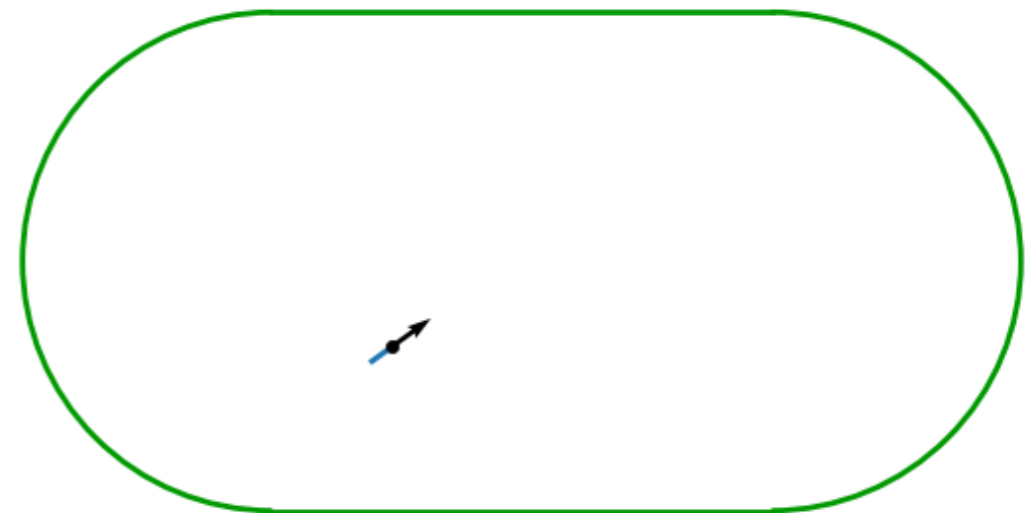
Google

Billiards in a “stadium” geometry



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Public transit



Wikipedia

UNPREDICTABILITY OF CHAOS

Classical Chaos: unpredictable deterministic evolution of the classical equations of motion.

Quantify chaos: Exponential sensitivity to initial conditions

$$\mathbf{x}'(t = 0) = \mathbf{x}(0) + \delta \quad \delta \ll 1$$

UNPREDICTABILITY OF CHAOS

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Quantify chaos: Exponential sensitivity to initial conditions

$$\mathbf{x}'(t=0) = \mathbf{x}(0) + \delta \quad \delta \ll 1$$

$$\|x(t) - x'(t)\| \sim e^{t\lambda_{\text{Lyapunov}}}$$

$$\lambda_{\text{Lyapunov}} > 0 \quad \text{Chaotic}$$

$$\lambda_{\text{Lyapunov}} < 0 \quad \text{Stable}$$

DYNAMICAL SYSTEMS

Continuous time $\frac{dx^1}{dt} = F_1(x^1, x^2, x^3, \dots, x^N)$

$$\frac{dx^2}{dt} = F_2(x^1, x^2, x^3, \dots, x^N)$$

\vdots

$$\frac{dx^N}{dt} = F_N(x^1, x^2, x^3, \dots, x^N)$$

N first order differential equations $\frac{d\mathbf{x}}{dt} = \mathbf{F}[\mathbf{x}(t)]$

Discrete time $\mathbf{x}_{n+1} = \mathbf{f}(\mathbf{x}_n)$

HAMILTONIAN SYSTEMS

Dynamics generated by a Hamiltonian depends on position, momentum, and time

$$H(\mathbf{q}, \mathbf{p}, t) \quad \mathbf{q} = (q_1, q_2, \dots, q_N)$$
$$\mathbf{p} = (p_1, p_2, \dots, p_N)$$

Equations of motion

$$d\mathbf{p}/dt = -\partial H(\mathbf{p}, \mathbf{q}, t)/\partial \mathbf{q},$$

$$d\mathbf{q}/dt = \partial H(\mathbf{p}, \mathbf{q}, t)/\partial \mathbf{p}.$$

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Equations of motion

$$\frac{d\mathbf{p}}{dt} = -\partial H(\mathbf{p}, \mathbf{q}, t) / \partial \mathbf{q}, \\ \frac{d\mathbf{q}}{dt} = \partial H(\mathbf{p}, \mathbf{q}, t) / \partial \mathbf{p}.$$

Written in the general construction, we have $\frac{d\tilde{\mathbf{x}}}{dt} = \mathbf{F}(\tilde{\mathbf{x}}, t) \quad \tilde{\mathbf{x}} = \begin{pmatrix} \mathbf{p} \\ \mathbf{q} \end{pmatrix}$
 $= \begin{pmatrix} d\mathbf{p}/dt \\ d\mathbf{q}/dt \end{pmatrix}$

HAMILTONIAN SYSTEMS

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$$\mathbf{F}(\tilde{\mathbf{x}}, t) = \mathbf{S}_N \cdot \partial H / \partial \tilde{\mathbf{x}} = \begin{pmatrix} d\mathbf{p}/dt \\ d\mathbf{q}/dt \end{pmatrix}$$

$$\frac{\partial H}{\partial \tilde{\mathbf{x}}} = \begin{bmatrix} \partial H / \partial \mathbf{p} \\ \partial H / \partial \mathbf{q} \end{bmatrix}$$

$$\mathbf{S}_N = \begin{bmatrix} \mathbf{O}_N & -\mathbf{I}_N \\ \mathbf{I}_N & \mathbf{O}_N \end{bmatrix}$$

NxN identity matrix

NxN zero matrix

ITERATIVE MAPS

Consider 1D maps $x_{n+1} = f(x_n)$

Initial condition $x_0 \xrightarrow{f} x_1 \xrightarrow{f} x_2 \xrightarrow{f} \cdots \xrightarrow{f} x_n$

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Weather forecasting



Tellus A:
Dynamic Meteorology and Oceanography

Reading: The problem of deducing the climate from the governing equations

Original Research Papers

The problem of deducing the climate from the governing equations

Edward D. Lorenz  (1964)

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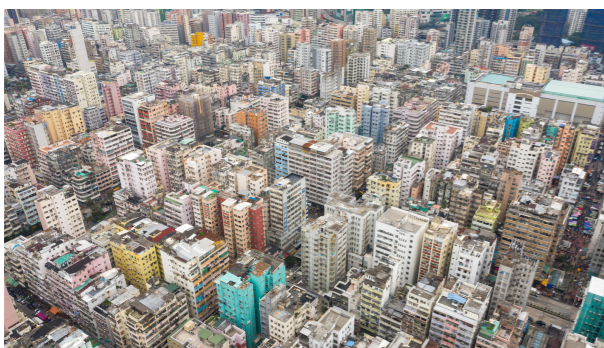
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Nature Vol. 261 June 10 1976

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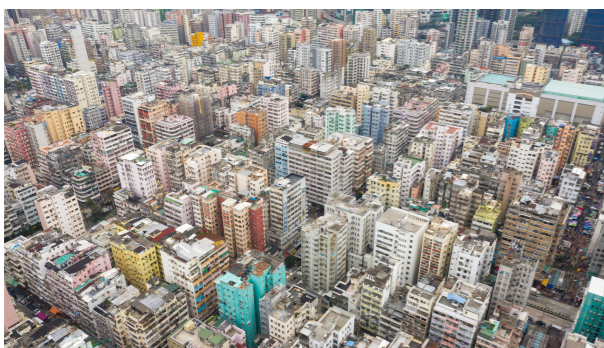
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Logistic map $x_{n+1} = rx_n(1 - x_n)$

FROM PERIODIC TO CHAOTIC

Logistic map

$$x_{n+1} = rx_n(1 - x_n)$$

Example: $r = 1.2$ **stable**

$$x_0 = 0.4$$

$$x'_0 = 0.4 + 10^{-9}$$

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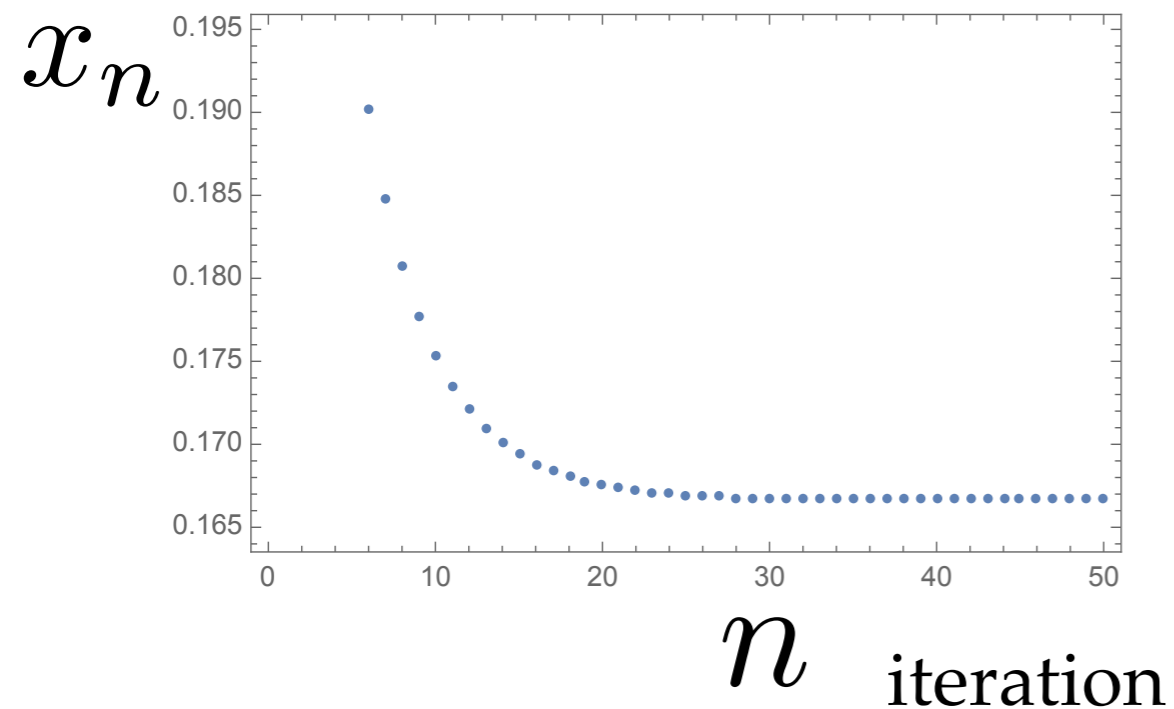
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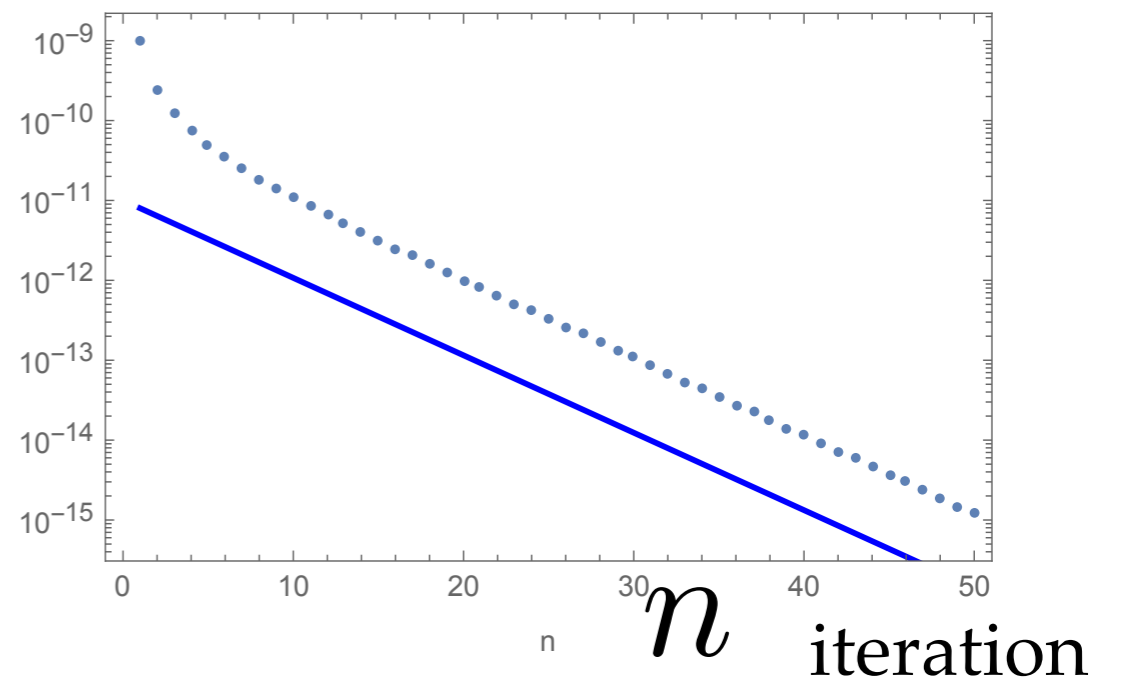
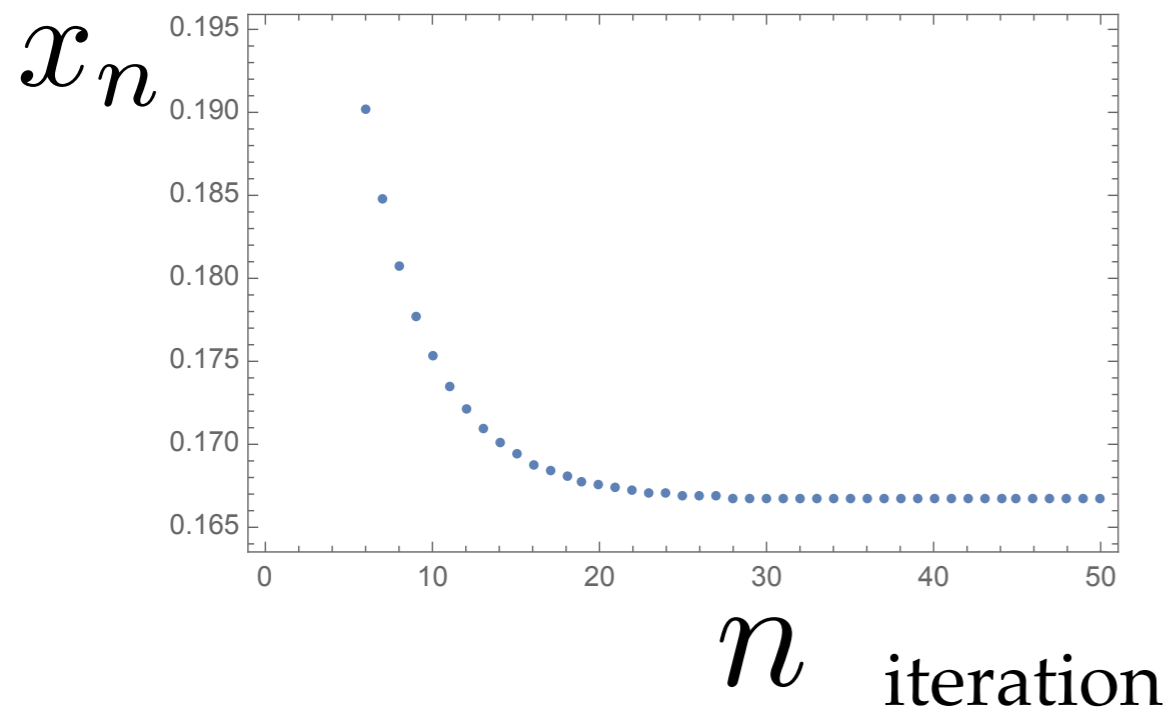
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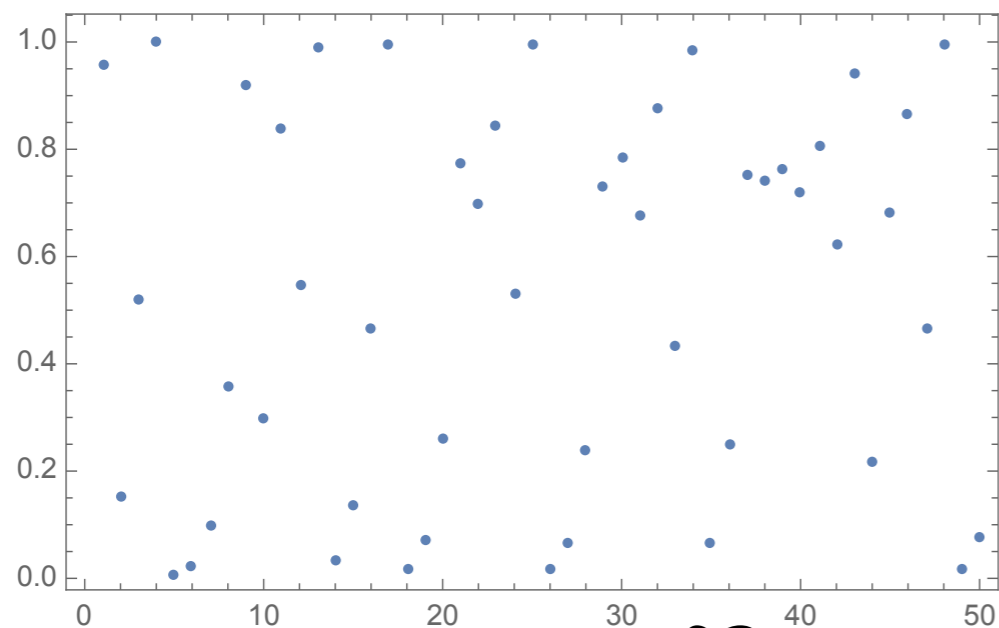
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x_n



n iteration

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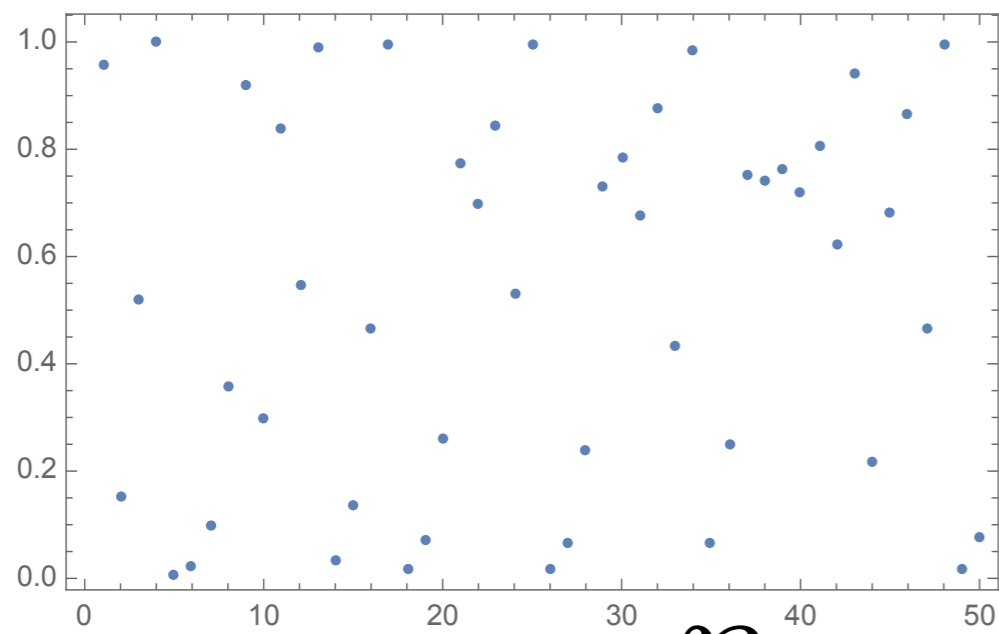
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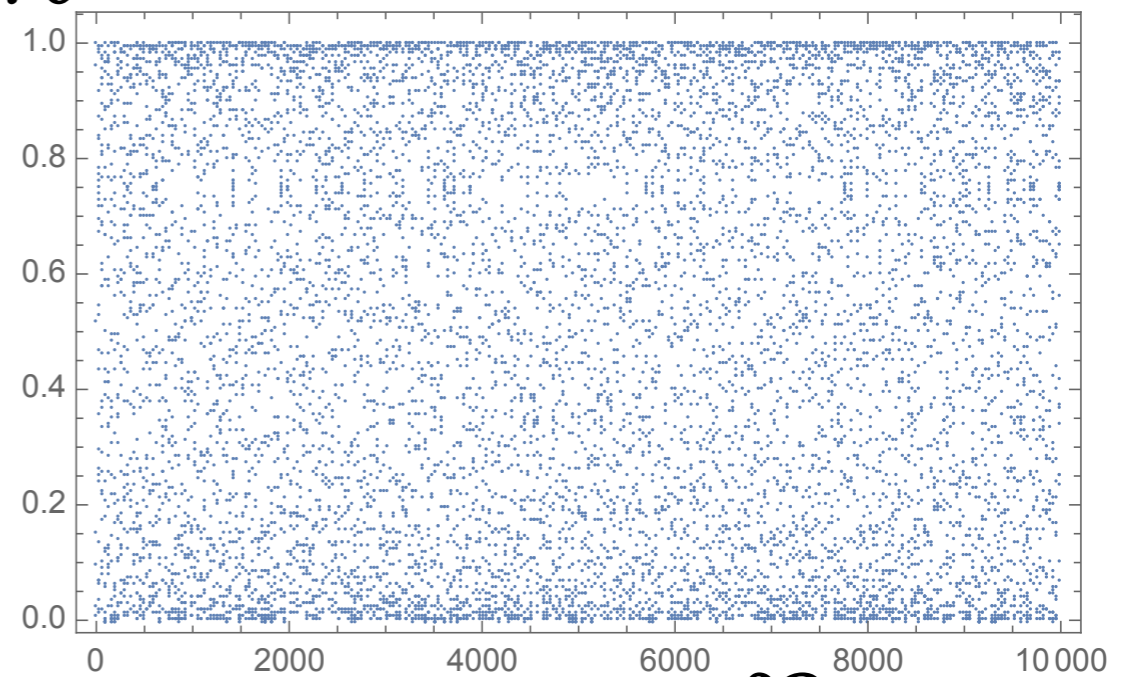
x_n



n iteration

x_n

Large number of iterations!



n iteration

FROM PERIODIC TO CHAOTIC

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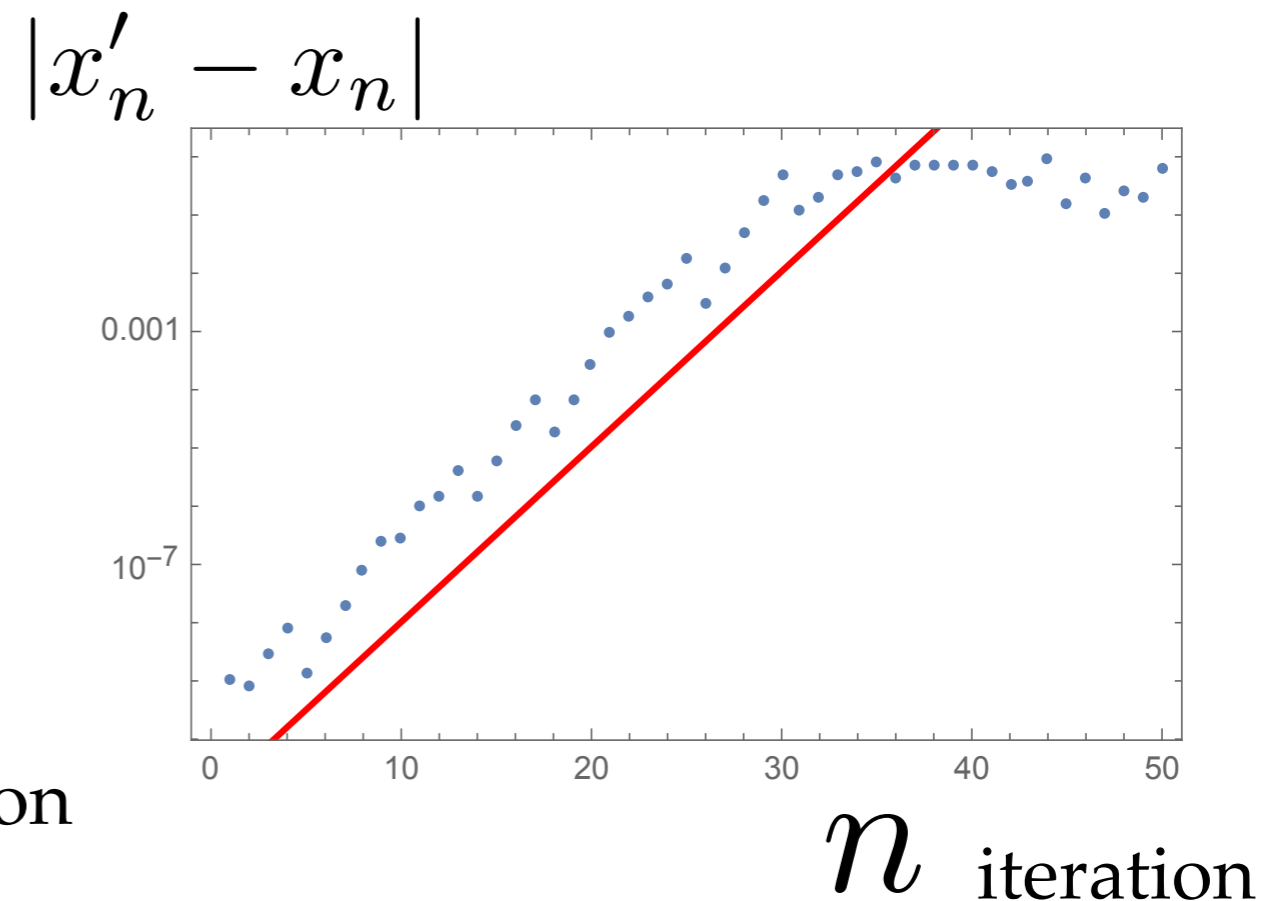
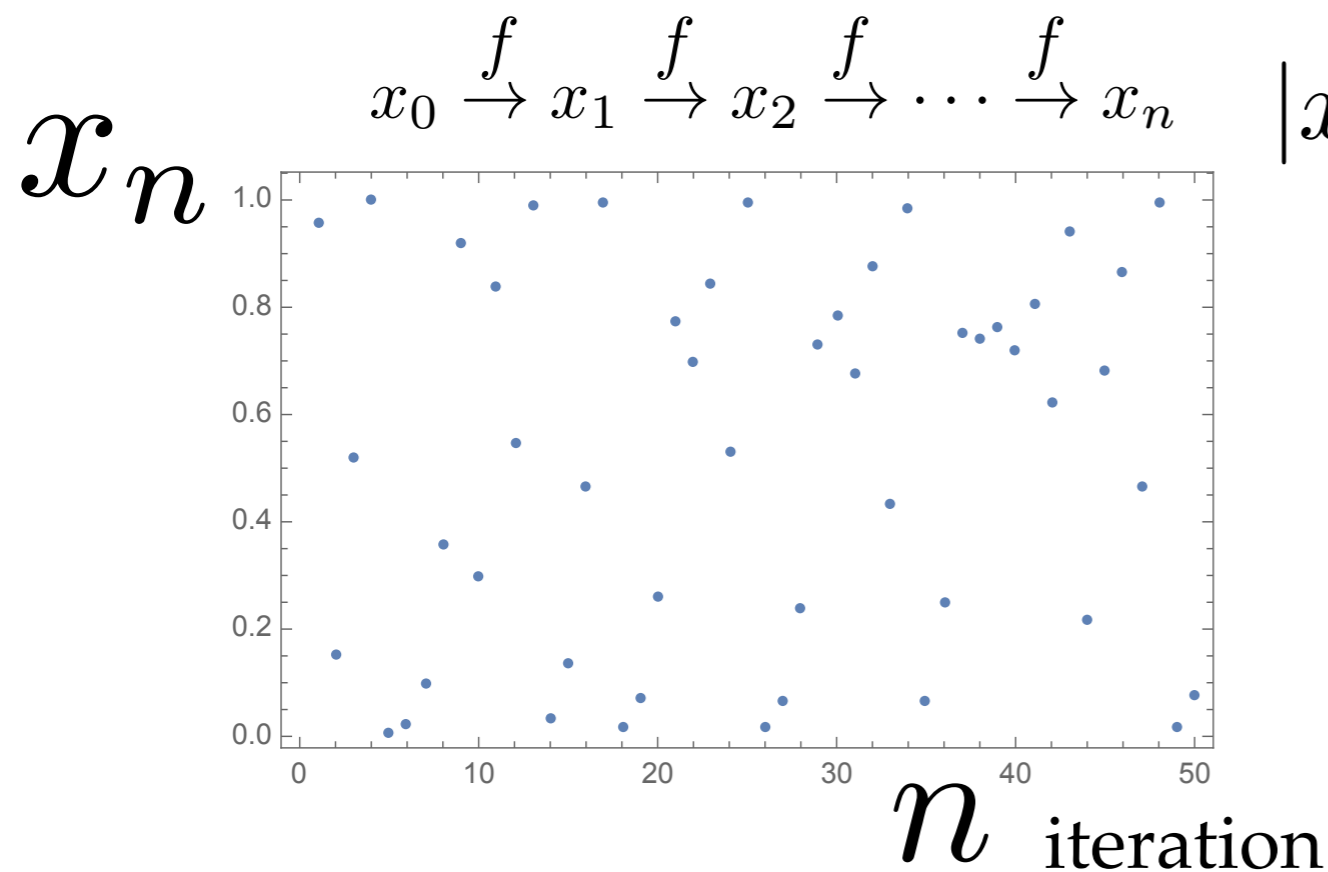
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THE LYAPUNOV EXPONENT

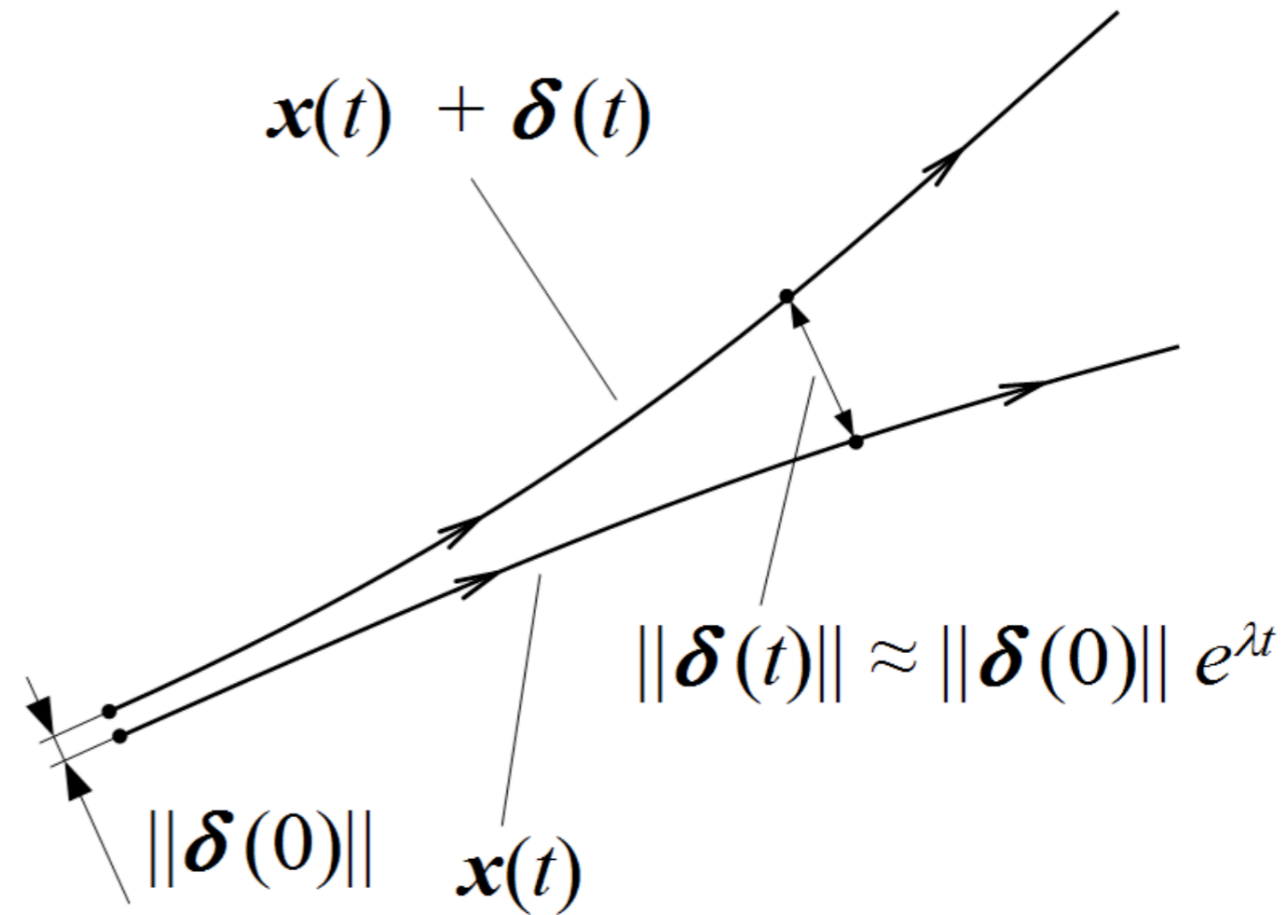
$$\mathbf{X}(t) = \mathbf{x}(t) + \delta(t)$$

Consider an infinitesimal perturbation

$$\frac{d\mathbf{x}}{dt} = \mathbf{F}[\mathbf{x}(t)]$$

Linearizing in δ

$$\frac{d\delta}{dt} = \frac{d\mathbf{F}}{d\mathbf{x}} \delta \equiv \mathbf{F}'[\mathbf{x}(t)] \delta(t)$$



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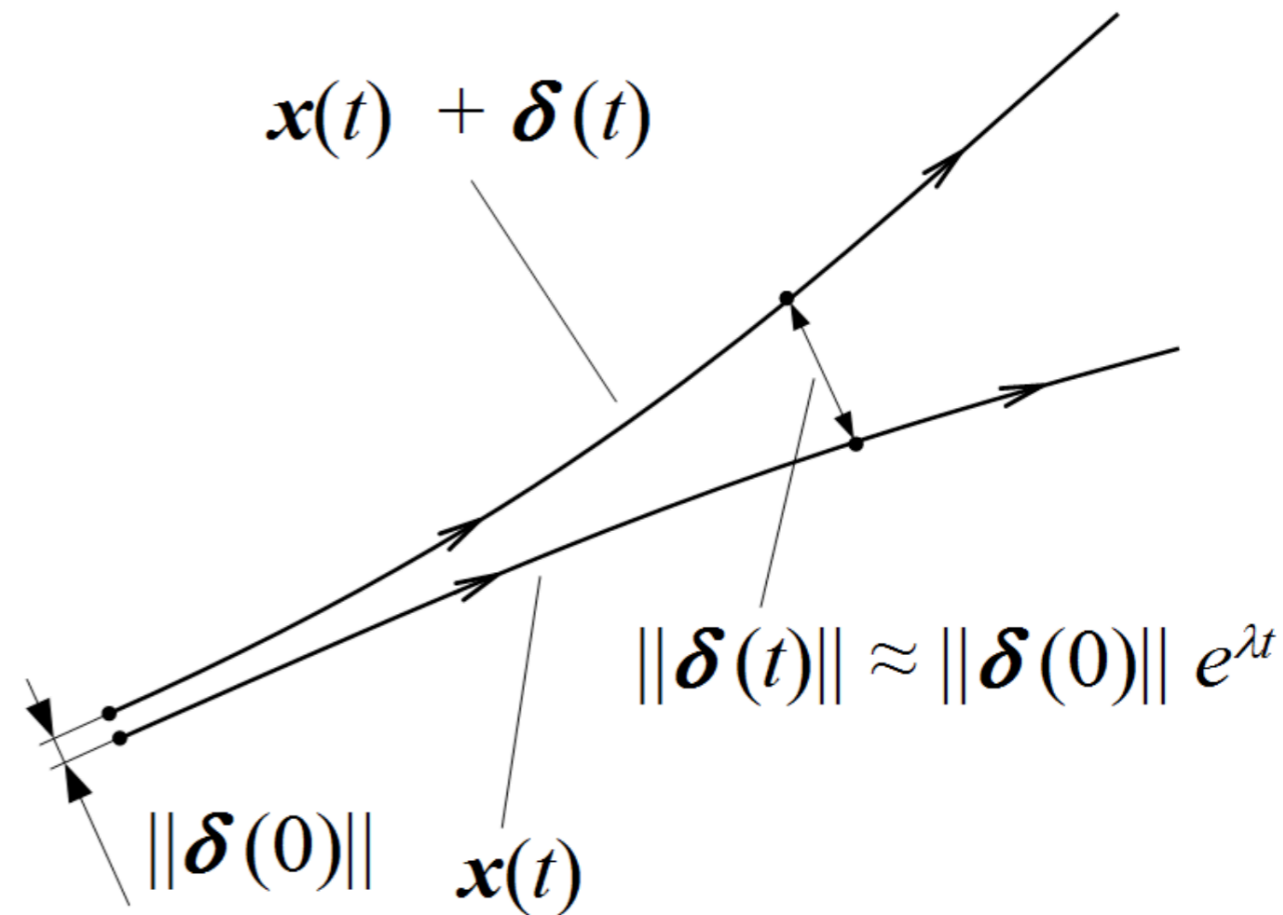
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$$\delta(t) = \exp \left\{ \int_0^t dt' \mathbf{F}'[\mathbf{x}(t')] \right\} \delta(0)$$



THE LYAPUNOV EXPONENT

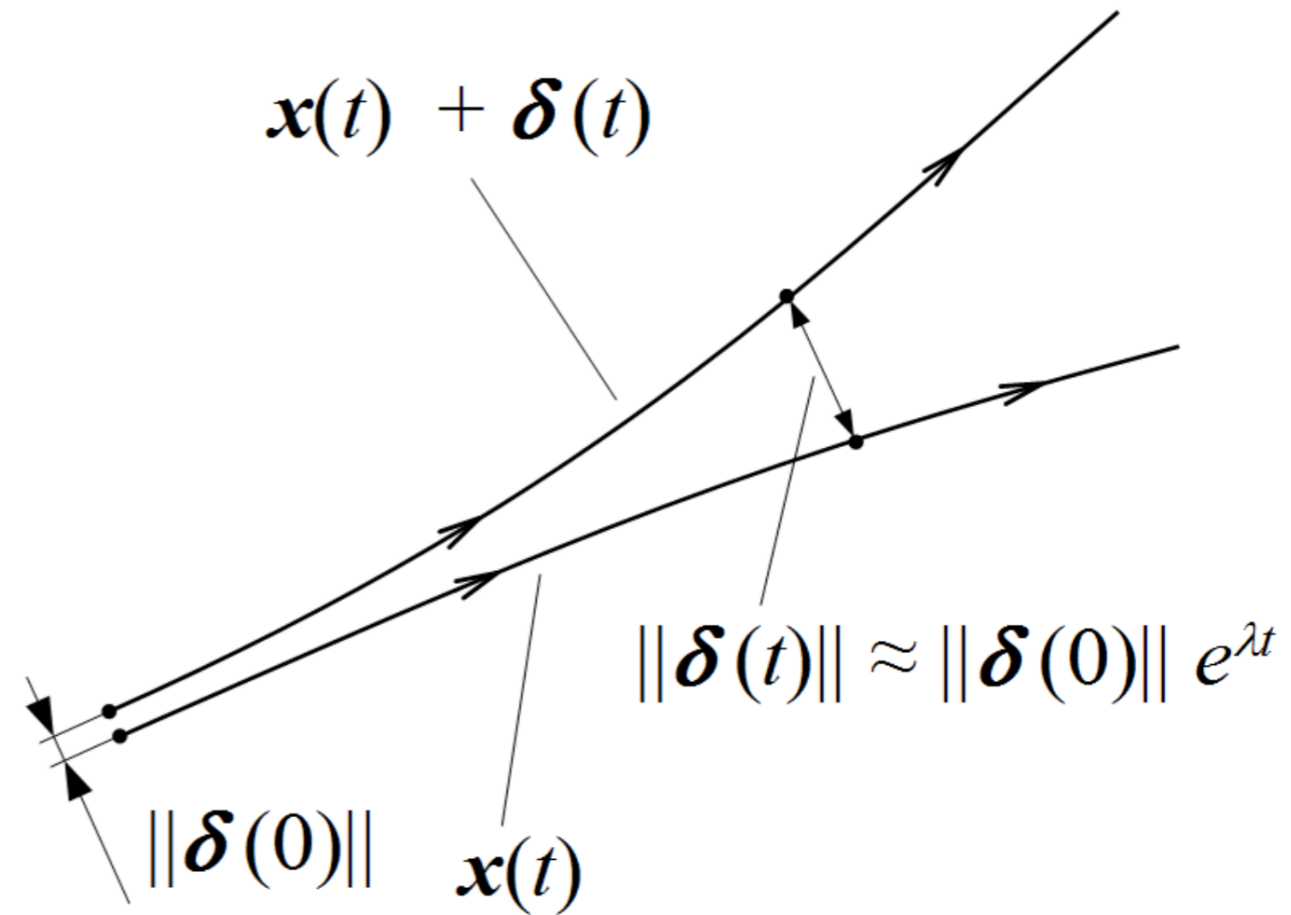
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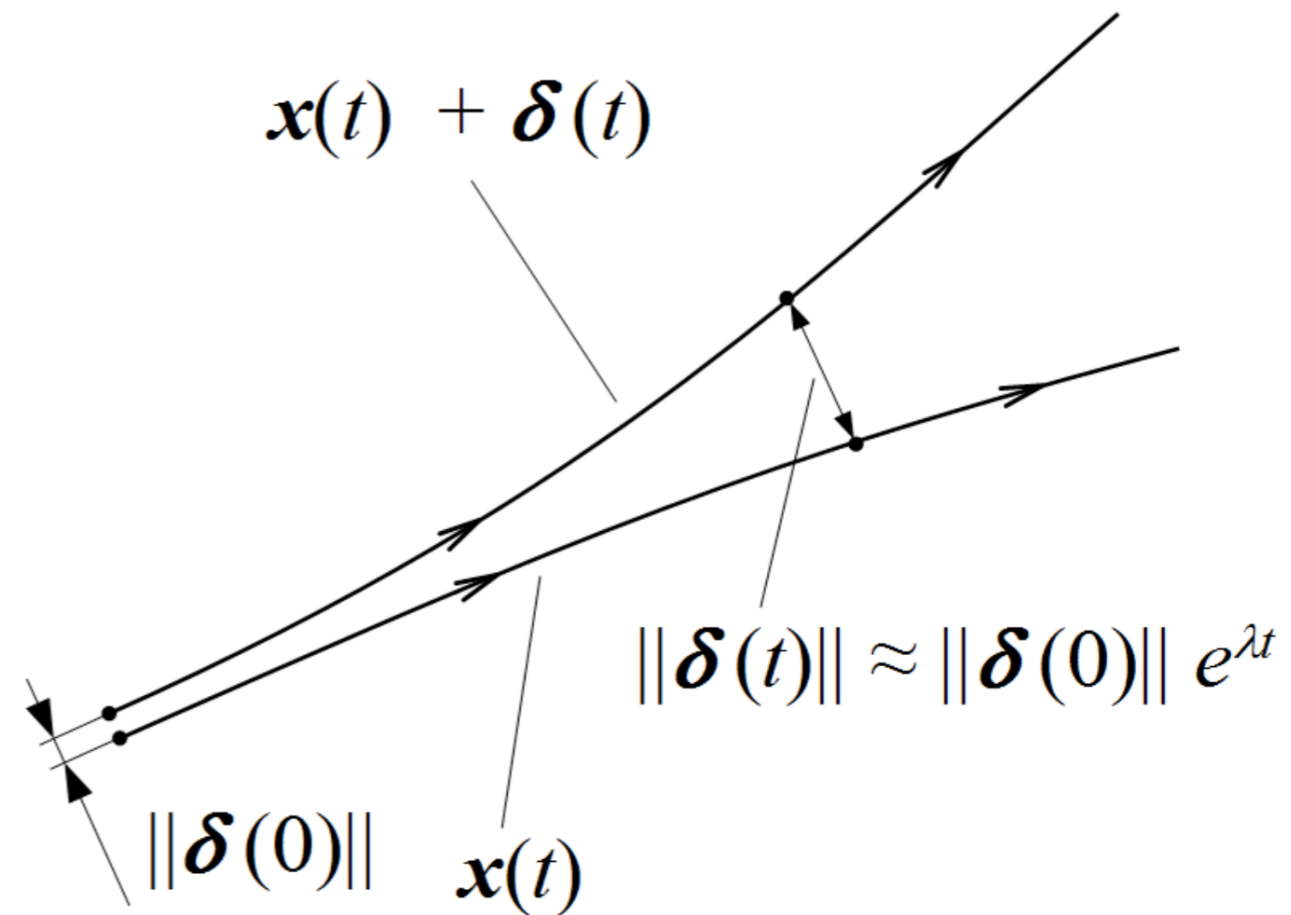
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$$\delta_n = \mathbf{f}'[\mathbf{x}_{n-1}] \mathbf{f}'[\mathbf{x}_{n-2}] \dots \mathbf{f}'[\mathbf{x}_0] \delta_0 = \prod_{k=0}^{n-1} \mathbf{f}'[\mathbf{x}_k] \delta_0$$



THE LYAPUNOV EXPONENT

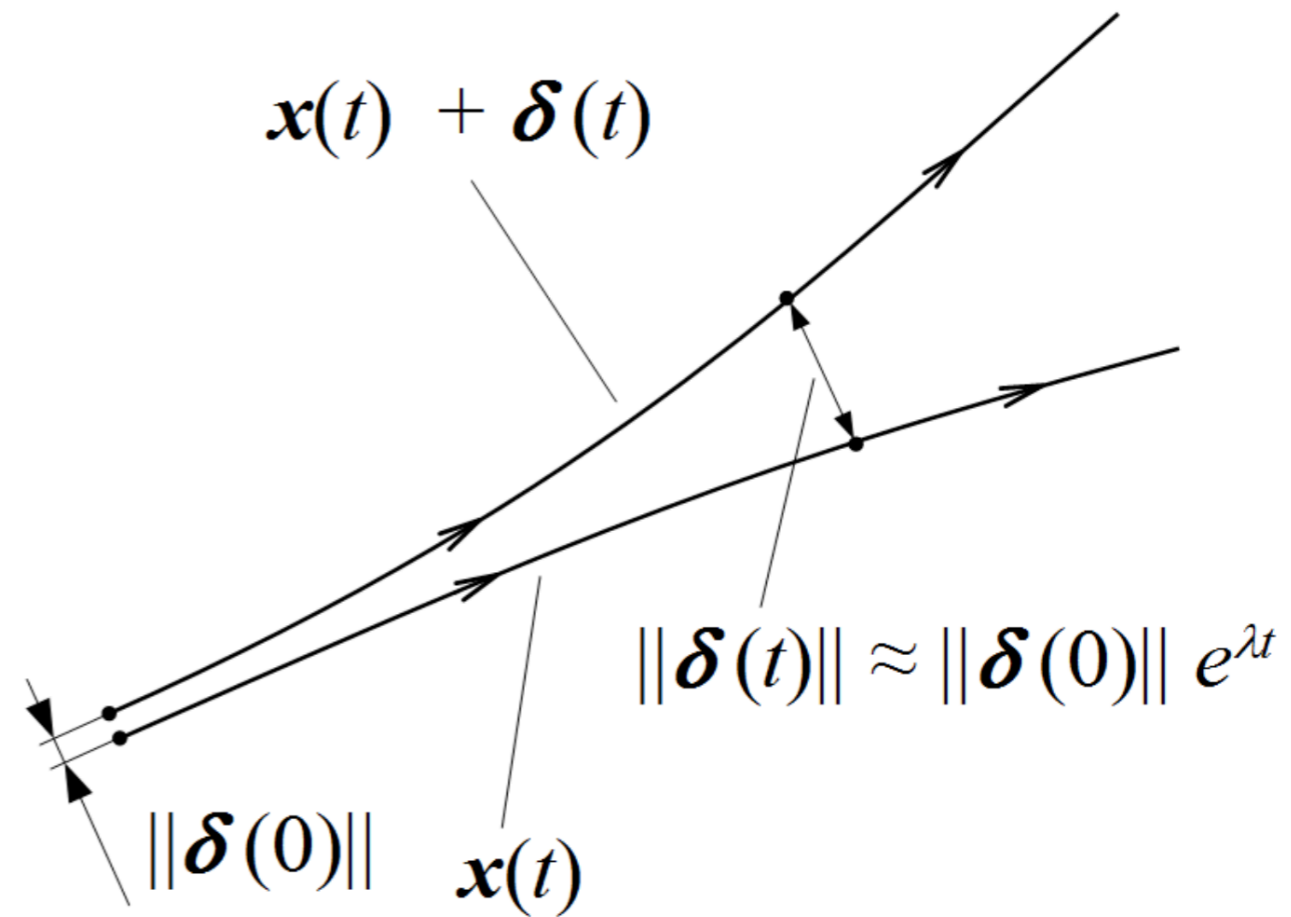
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$$\log |\delta_n| = \sum_{k=0}^{n-1} \log |\mathbf{f}'[\mathbf{x}_k]| + \log |\delta_0|$$



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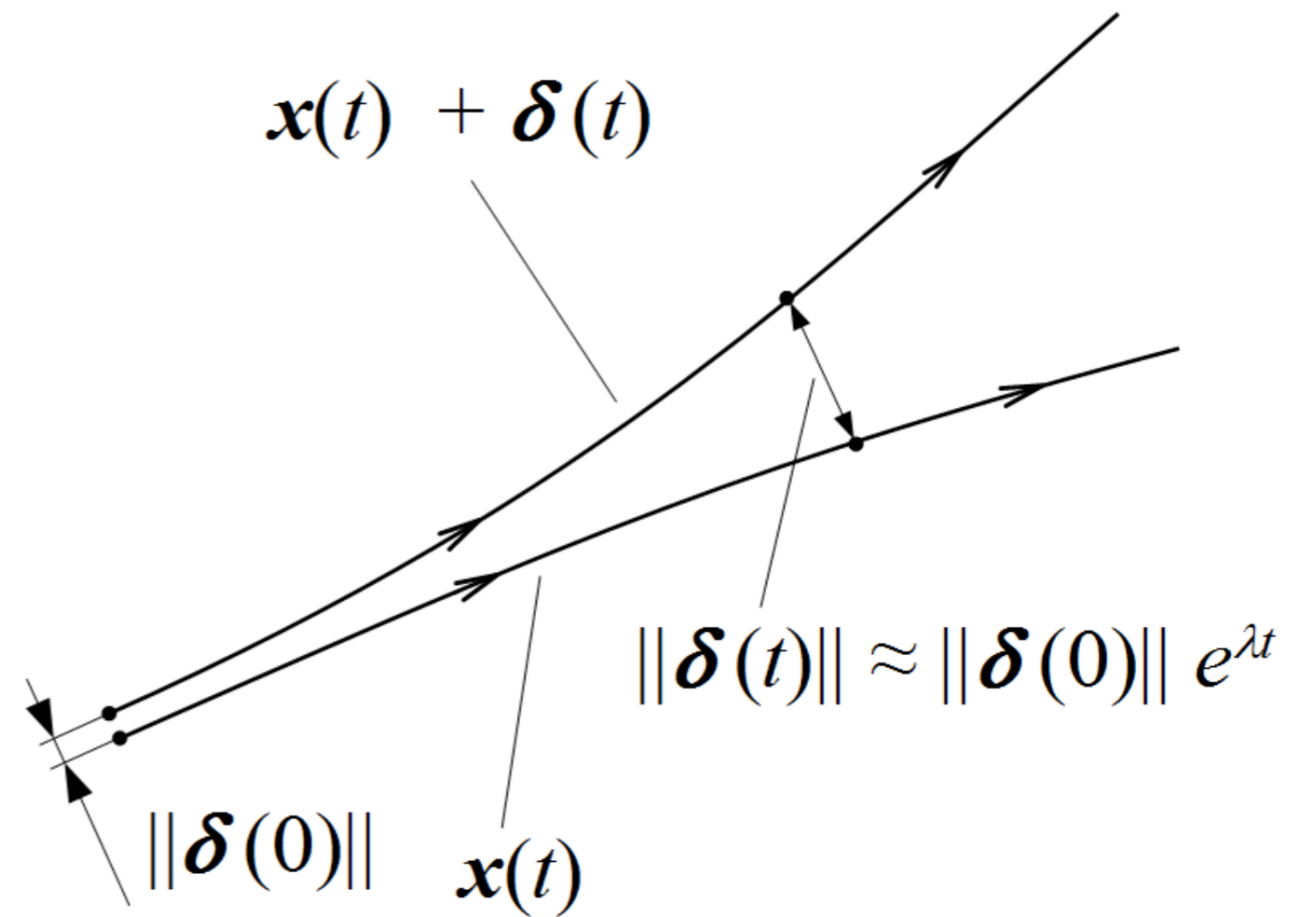
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THE LYAPUNOV EXPONENT

$$\delta(t) = \exp \left\{ \int_0^t dt' \mathbf{F}'[\mathbf{x}(t')] \right\} \delta(0) \sim e^{\lambda t}$$

The Lyapunov Exponent
(continuous time)

$$\lambda = \lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t dt' \mathbf{F}'[\mathbf{x}(t')]$$

$$|\delta_n| = |\delta_0| \exp \left(\sum_{k=0}^{n-1} \log |\mathbf{f}'[\mathbf{x}_k]| \right) \sim e^{\lambda n}$$

The Lyapunov Exponent
(discrete time)

$$\lambda = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} \log |\mathbf{f}'[\mathbf{x}_k]|$$

THE LYAPUNOV EXPONENT

Logistic map

$$x_{n+1} = rx_n(1 - x_n)$$

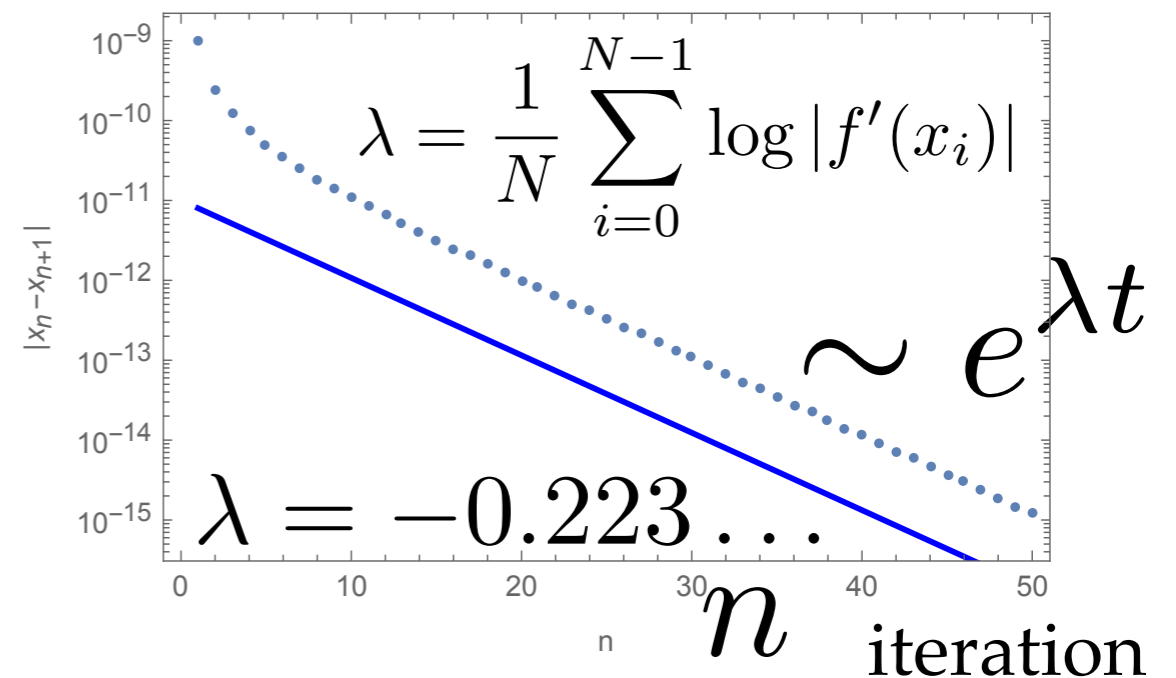
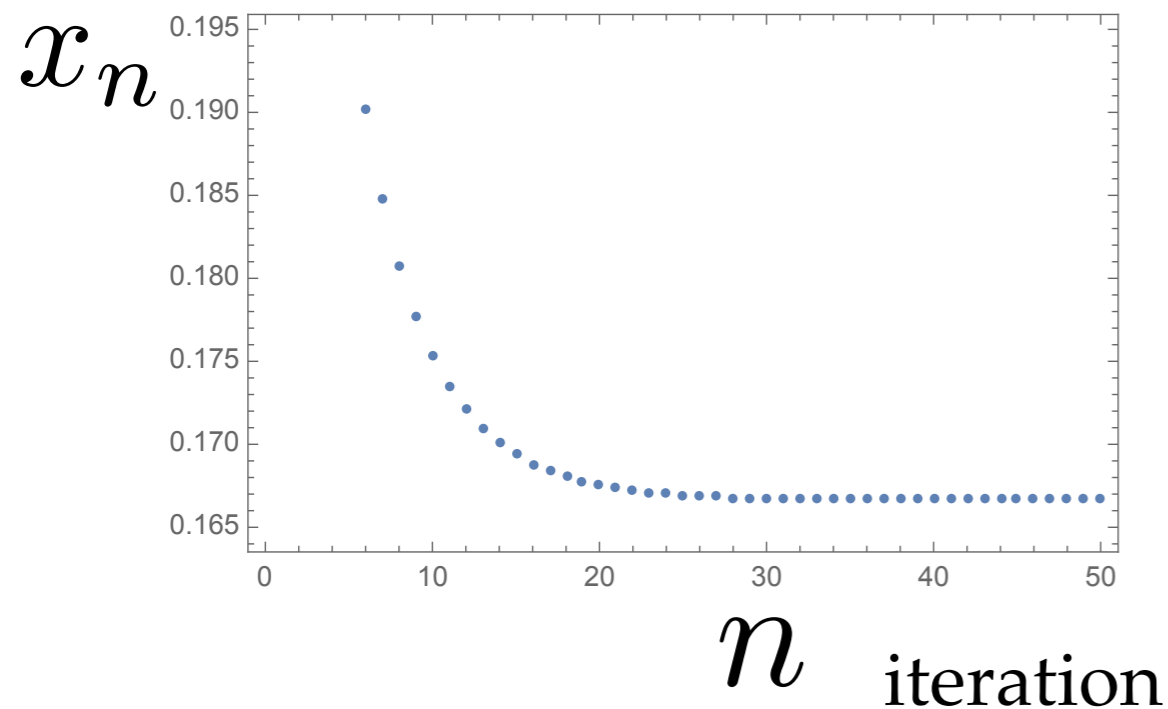
Example: $r = 1.2$ **stable**

$$x_0 = 0.4$$

$$x'_0 = 0.4 + 10^{-9} \quad x'_0 \xrightarrow{f} x'_1 \xrightarrow{f} x'_2 \xrightarrow{f} \dots \xrightarrow{f} x'_n$$

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$$|x'_n - x_n|$$



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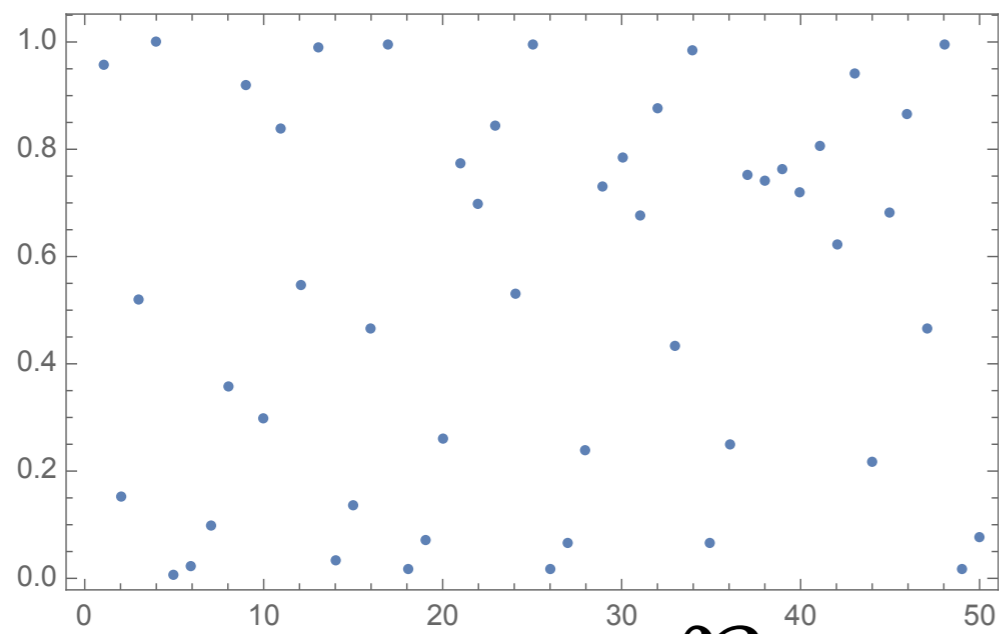
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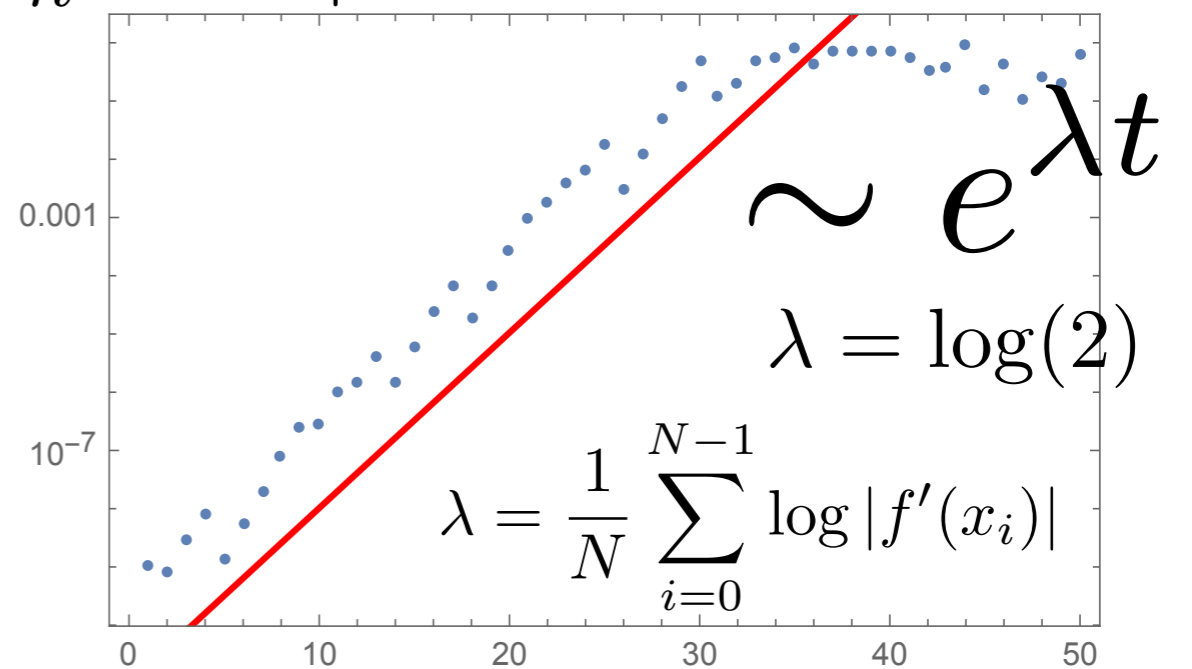
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x_n



$$|x'_n - x_n|$$



n iteration

ROUTE TO CHAOS: MAPS

Logistic map

$$x_{n+1} = rx_n(1 - x_n)$$

Vary parameter, r

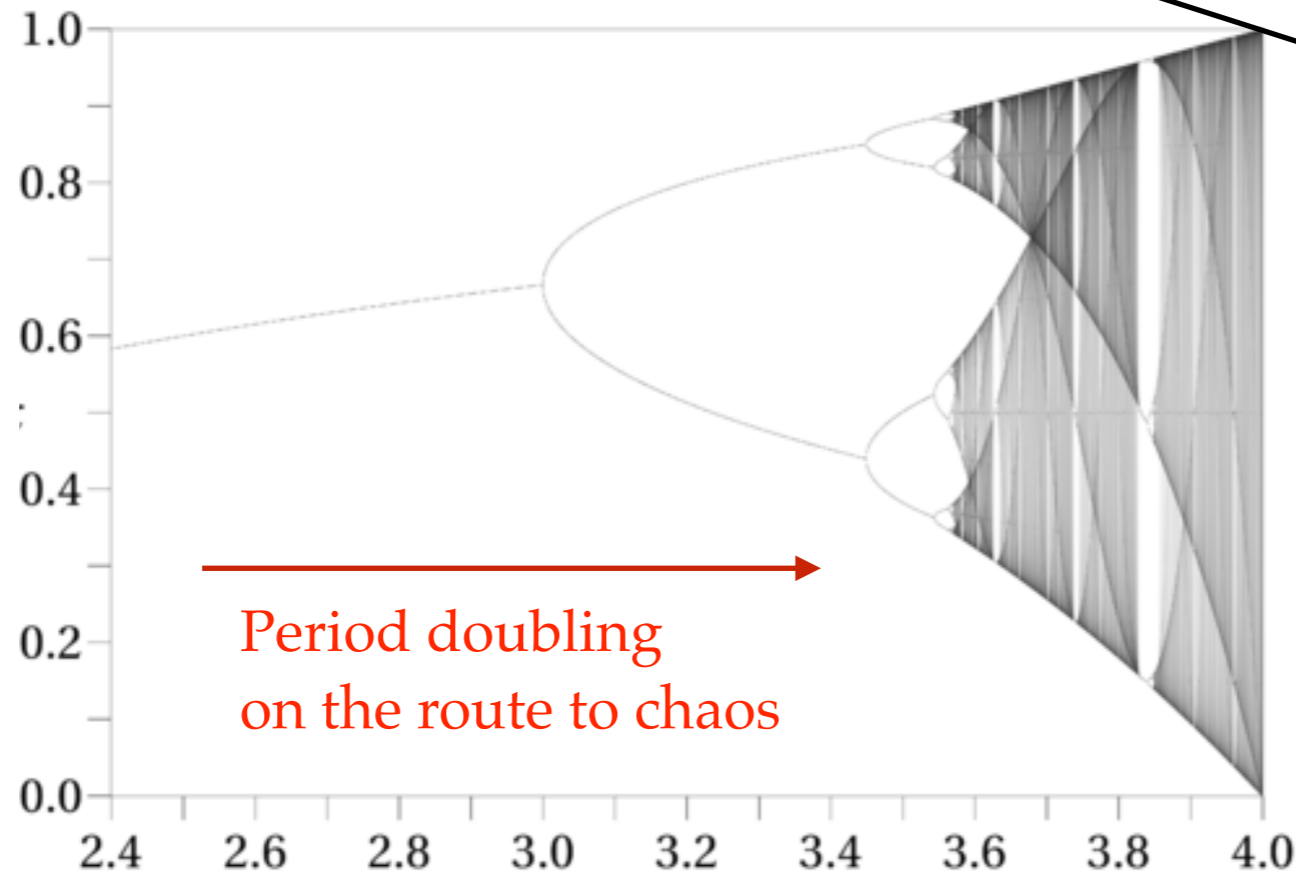
in the limit $n \rightarrow \infty$

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$\lim_{n \rightarrow \infty} x_n$



Vary parameter, r
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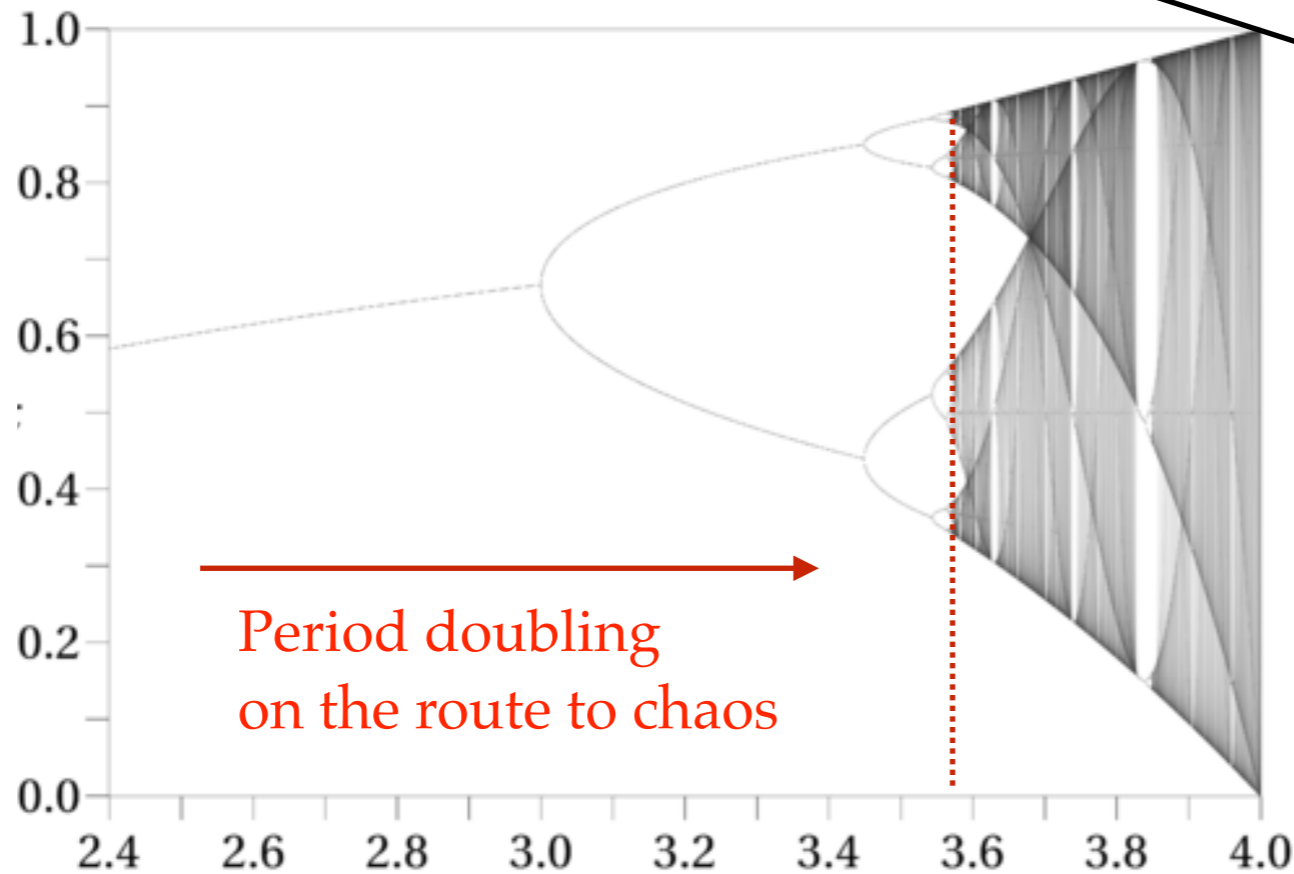
"Strength of chaos" r

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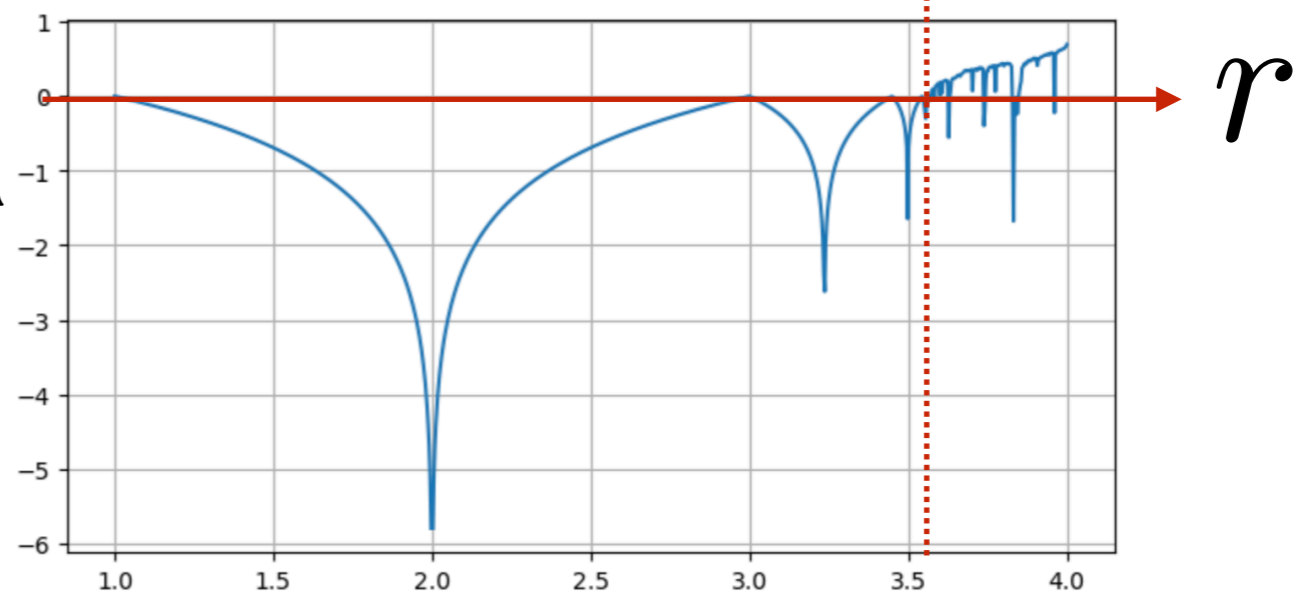


Vary parameter, r

Period doubling
on the route to chaos

"Strength of chaos" r

Lyapunov
exponent λ

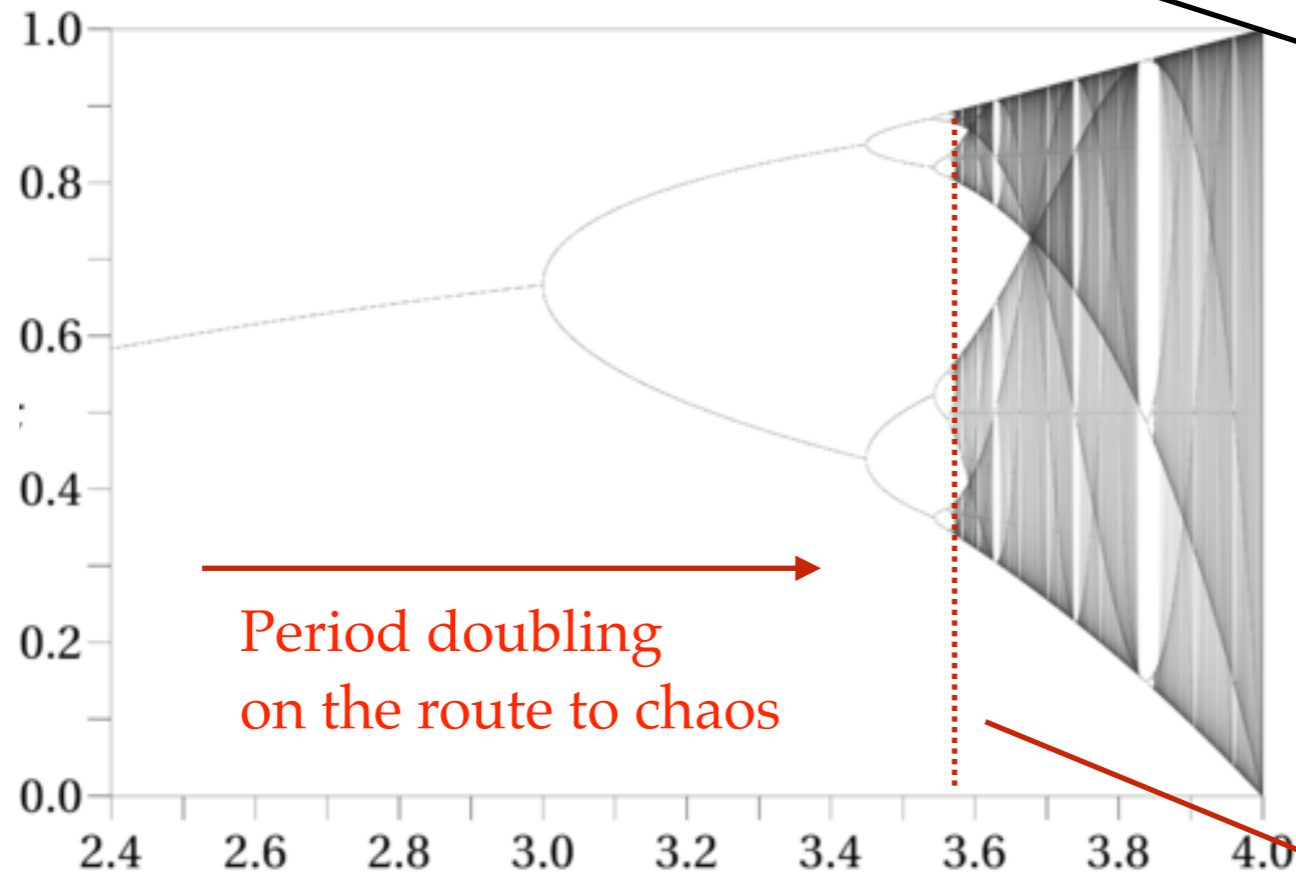


ROUTE TO CHAOS: MAPS

Logistic map

$$x_{n+1} = rx_n(1 - x_n)$$

$\lim_{n \rightarrow \infty} x_n$



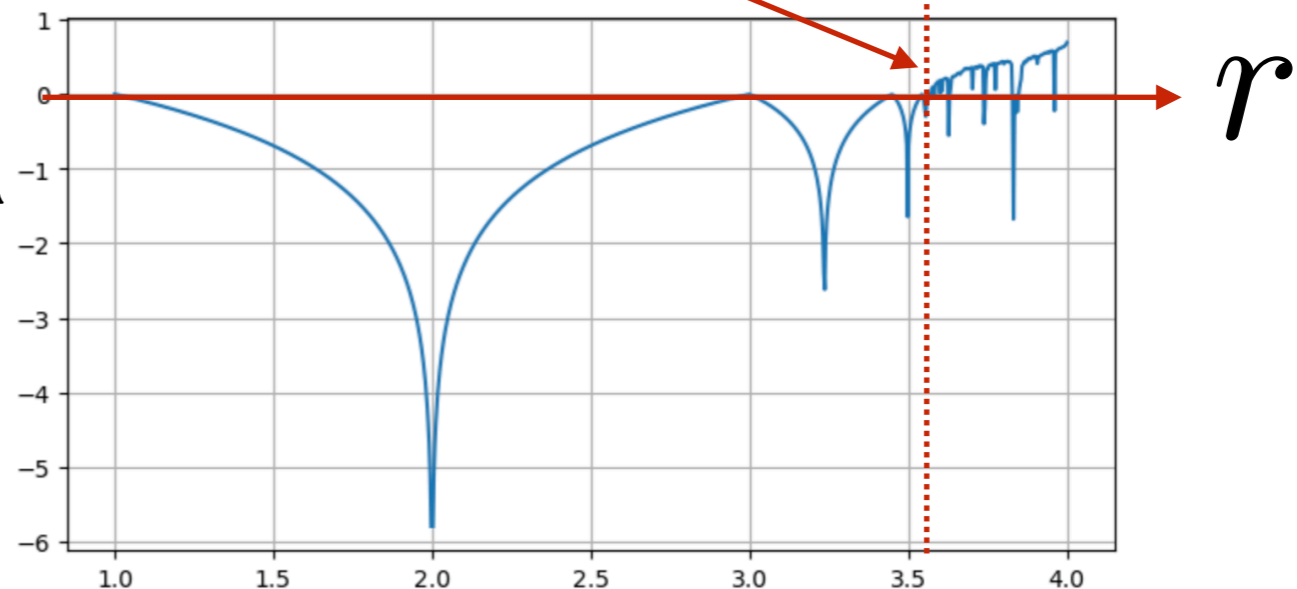
Vary parameter, r

Period doubling
on the route to chaos

Onset of chaos

"Strength of chaos" r

Lyapunov
exponent λ



A SIMPLER CHAOTIC MAP

For $r=4$, where this is chaotic it can be reduced to the Bernoulli map

$$x_{n+1} = 4x_n(1 - x_n) \qquad x_n = \sin^2\left(\frac{\pi t_n}{2}\right)$$

$$\sin^2(\pi t_{n+1}/2) = 4 \sin^2(\pi t_n/2) \cos^2(\pi t_n/2) = \sin^2(\pi t_n)$$

$$\longrightarrow \pi t_{n+1}/2 = \pi t_n \pmod{\pi}$$

Bernoulli map
(or doubling map)

$$t_{n+1} = 2t_n \pmod{1}$$

$$0 < t < 1$$

$$\lambda_{\text{Lyapunov}} = \ln 2$$

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binary bit strings on a classical computer

$$t_n = 0.b_1b_2b_3 \dots$$

$$b_i = 0, 1$$

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Starting from

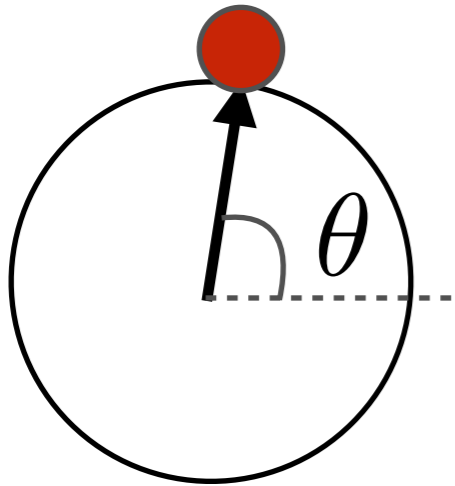
$$t_0 = \frac{1}{\sqrt{2}}$$



CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

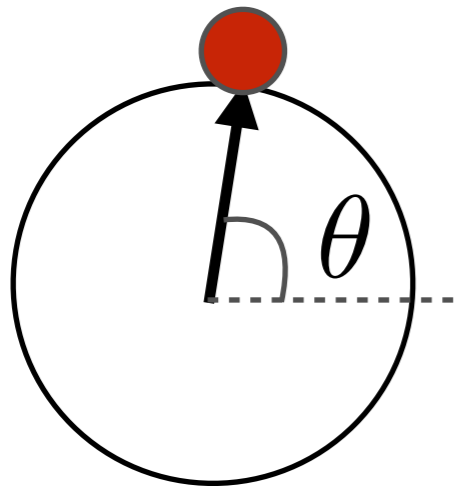
Kicked rotor

$$H_{\text{KR}} = \frac{p^2}{2I} + k \cos(\theta) \sum_{n=-\infty}^{+\infty} \delta(t - nT)$$



CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

Kicked rotor



$$H_{\text{KR}} = \frac{p^2}{2I} + k \cos(\theta) \sum_{n=-\infty}^{+\infty} \delta(t - nT)$$

Equations of motion

$$\frac{dp}{dt} = k \sin \theta \sum_n \delta(t - nT)$$

$$\frac{d\theta}{dt} = \frac{p}{I}$$

Constant at times
not equal to nT

Let p_n, θ_n be these
constant at times $t = nT + 0^+$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

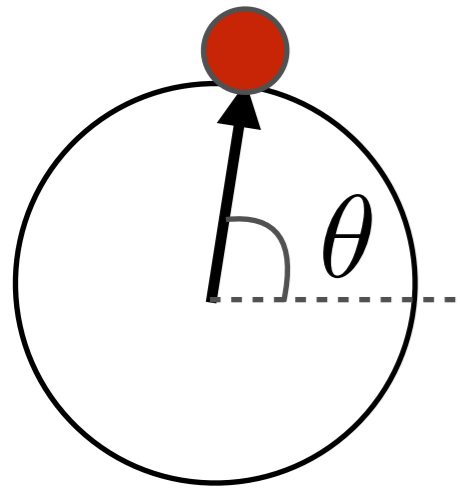
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Integrating the equations,
across one period

$$\int_{mT}^{(m+1)T} dt(\dots)$$

Let p_n, θ_n be these
constant at times $t = nT + 0^+$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

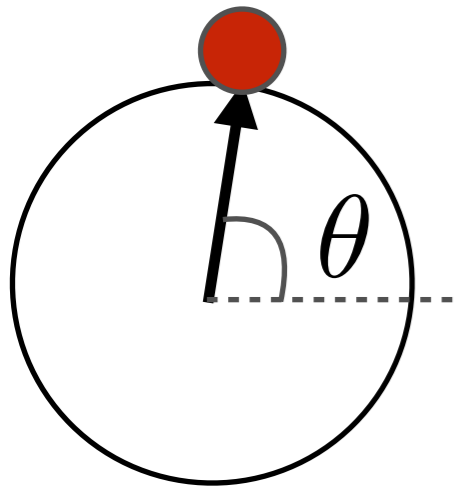
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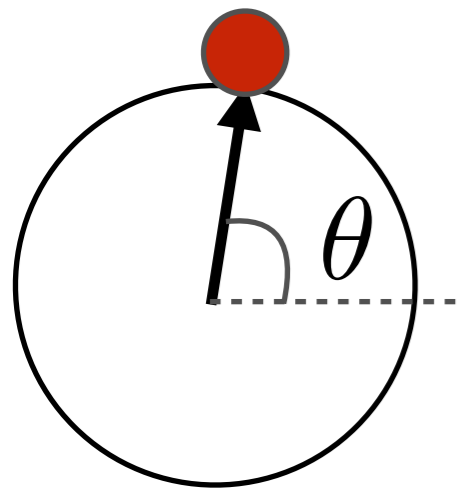
$$p((m+1)T) - p(mT) = p_{m+1} - p_m = k \sin \theta_{m+1}$$

$$\theta((m+1)T) - \theta(mT) = \theta_{m+1} - \theta_m = p_m \frac{T}{I}$$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

Kicked rotor

This is the “standard map”



$$\theta_{m+1} = \left(\theta_m + p_m \frac{T}{I} \right) \text{ modulo } 2\pi$$

Setting $T/I=1$

$$p_{m+1} = k \sin \theta_{m+1} + p_m$$

$k = 0$
Integrable

$$p_{m+1} = p_m$$

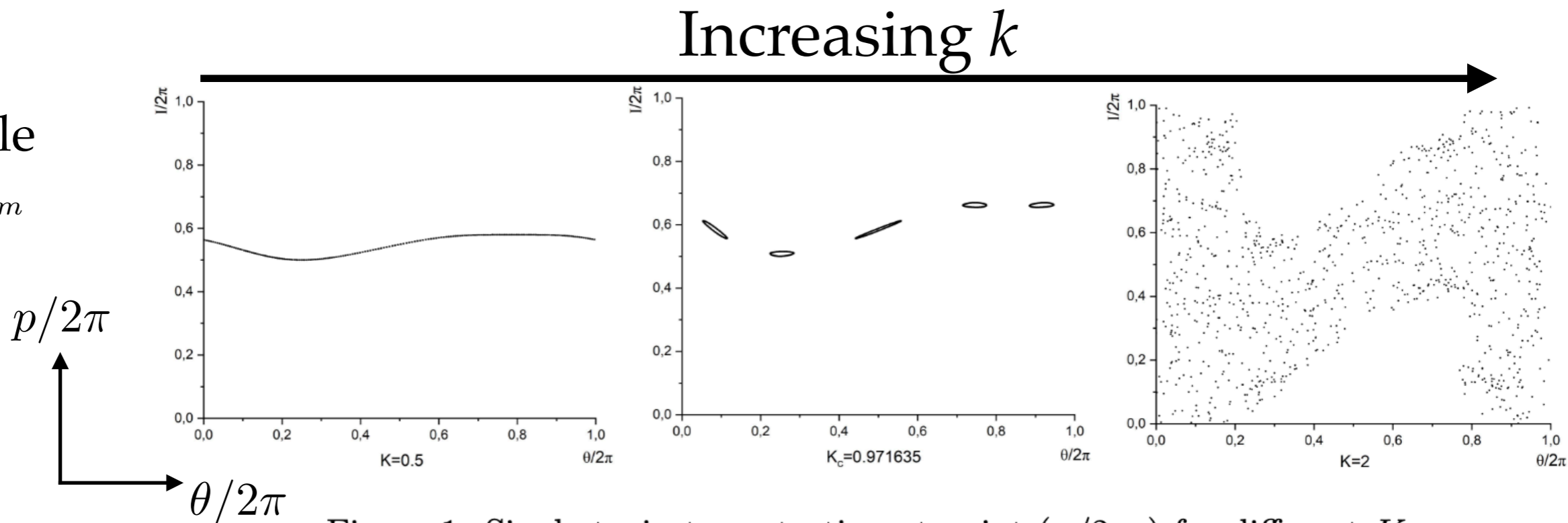
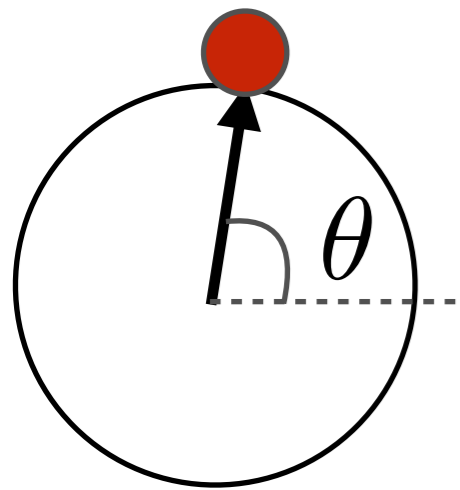


Figure 1: Single trajectory starting at point $(\pi/2, \pi)$ for different K

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Setting $T/I=1$

$$p_{m+1} = k \sin \theta_{m+1} + p_m$$

Mixed phase space

Critical

Chaotic

$$k = 0$$

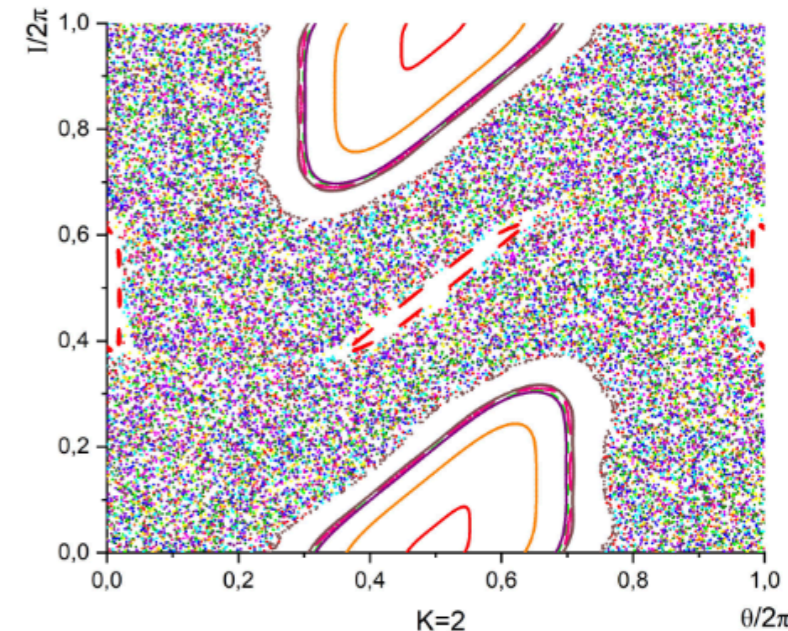
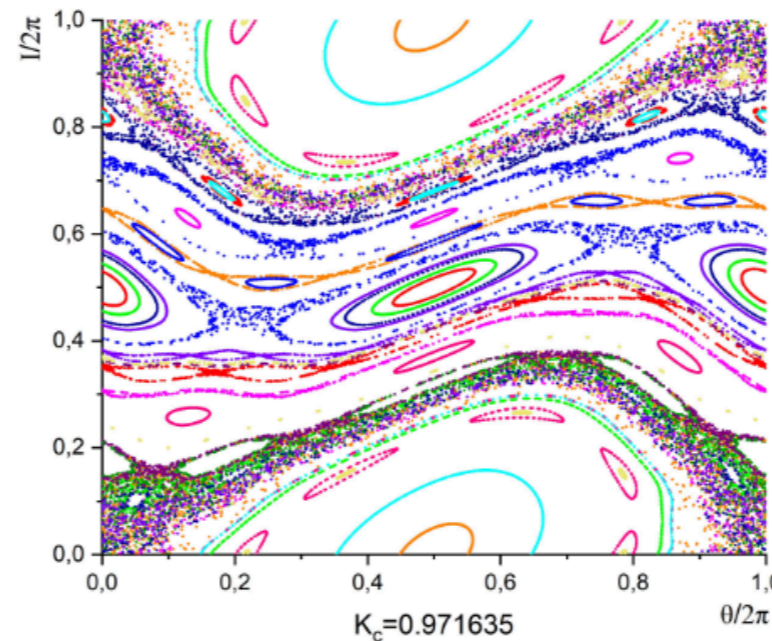
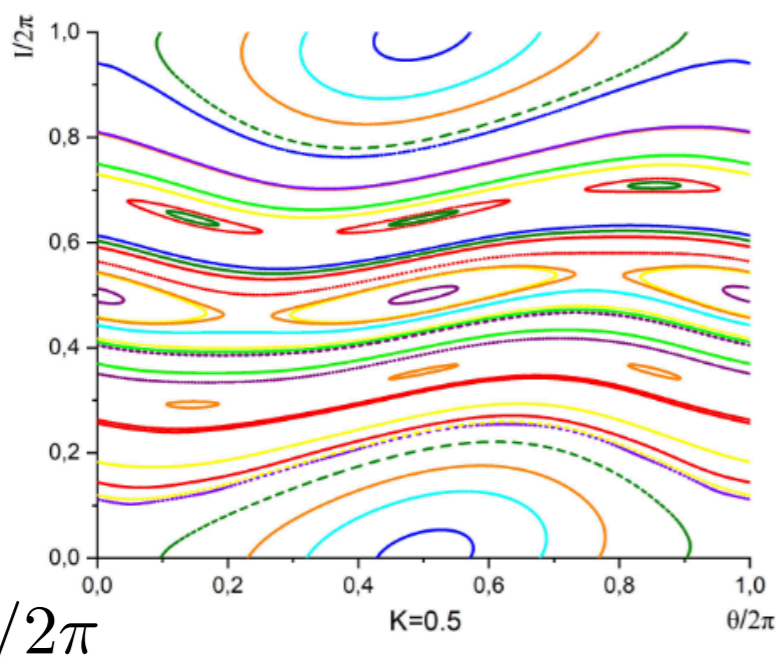
$$k < k_c$$

$$k_c = 0.971635$$

$$k > k_c$$

Integrable

$$p_{m+1} = p_m$$



$$p/2\pi$$

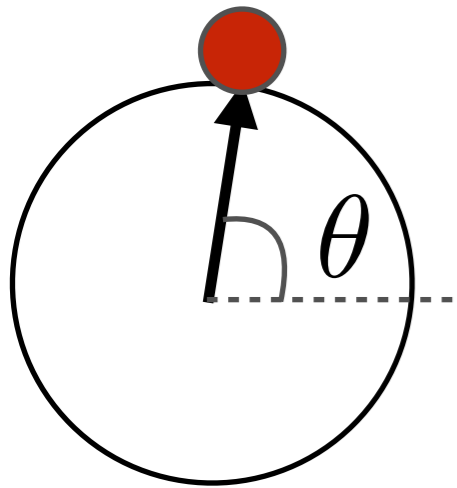
$$\theta/2\pi$$

Figure 2: Poincaré surfaces for different K

CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

Kicked rotor

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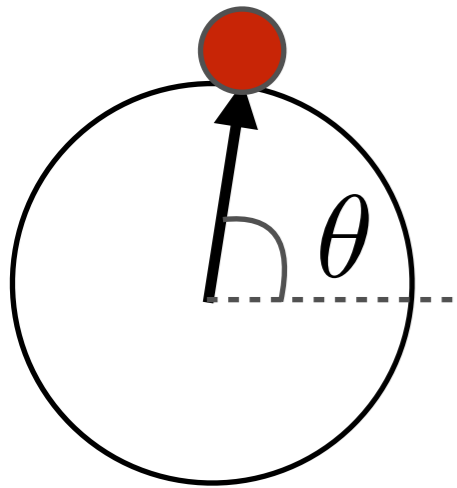
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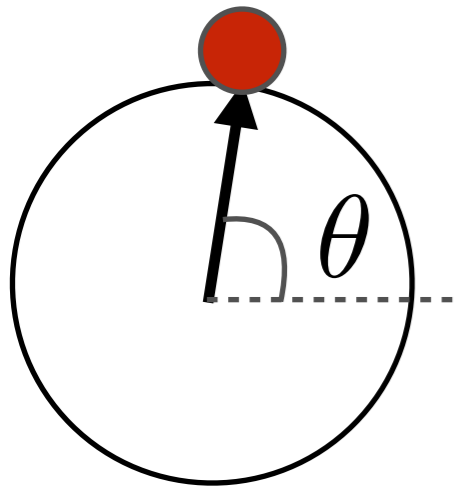
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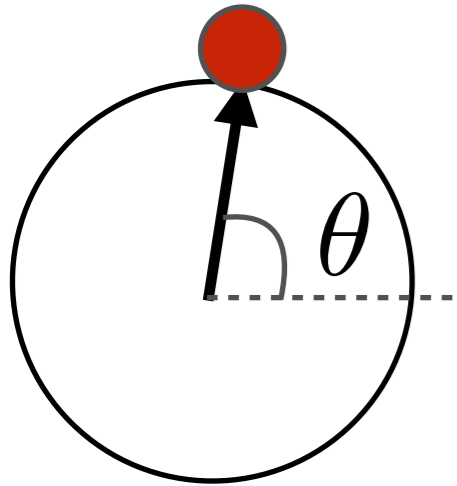
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$$p_{m+1} = \sum_{j=1}^{m+1} \sin(\theta_j) + p_0$$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

Kicked rotor



Diffusion in the kicked rotor

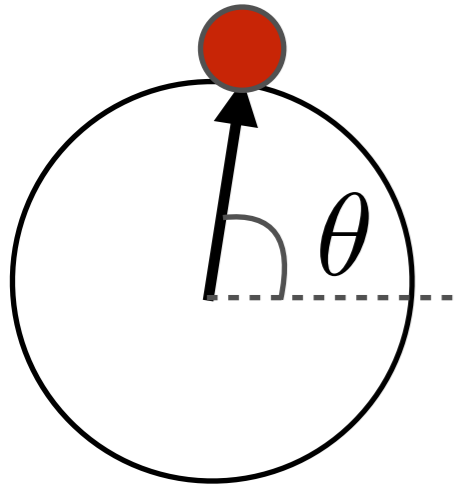
$$p_{m+1} = \sum_{j=1}^{m+1} \sin(\theta_j) + p_0$$

Average over kicks

$$\langle (p_m - p_0)^2 \rangle = k^2 \sum_{i,j}^m \langle \sin \theta_i \sin \theta_j \rangle$$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

Kicked rotor



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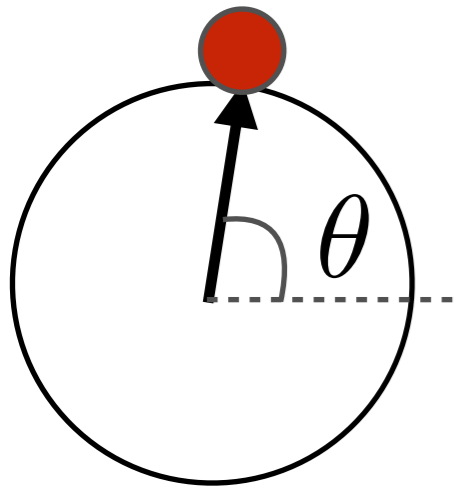
$$\langle (p_m - p_0)^2 \rangle = k^2 \sum_{i,j}^m \langle \sin \theta_i \sin \theta_j \rangle$$

In the limit $k \gg 1$, kicks are totally random and uncorrelated

$$\langle \sin \theta_i \sin \theta_j \rangle = \delta_{i,j} \langle \sin^2(\theta_i) \rangle = \delta_{i,j} \frac{1}{2}$$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

Kicked rotor



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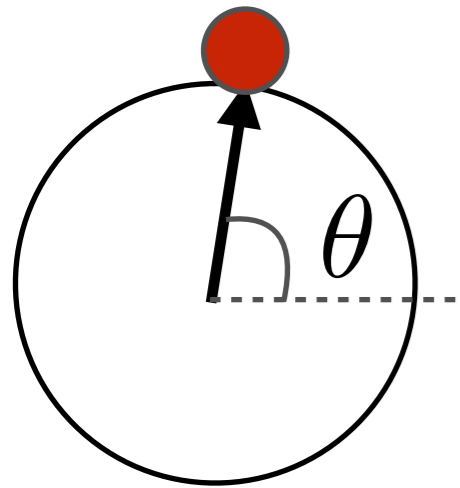
$$\langle (p_m - p_0)^2 \rangle = k^2 m / 2$$

The root-mean-square of a diffusion process

$$\langle (p_m - p_0)^2 \rangle_{\text{rms}} = 2Dm \quad D \approx k^2 / 4$$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

Kicked rotor

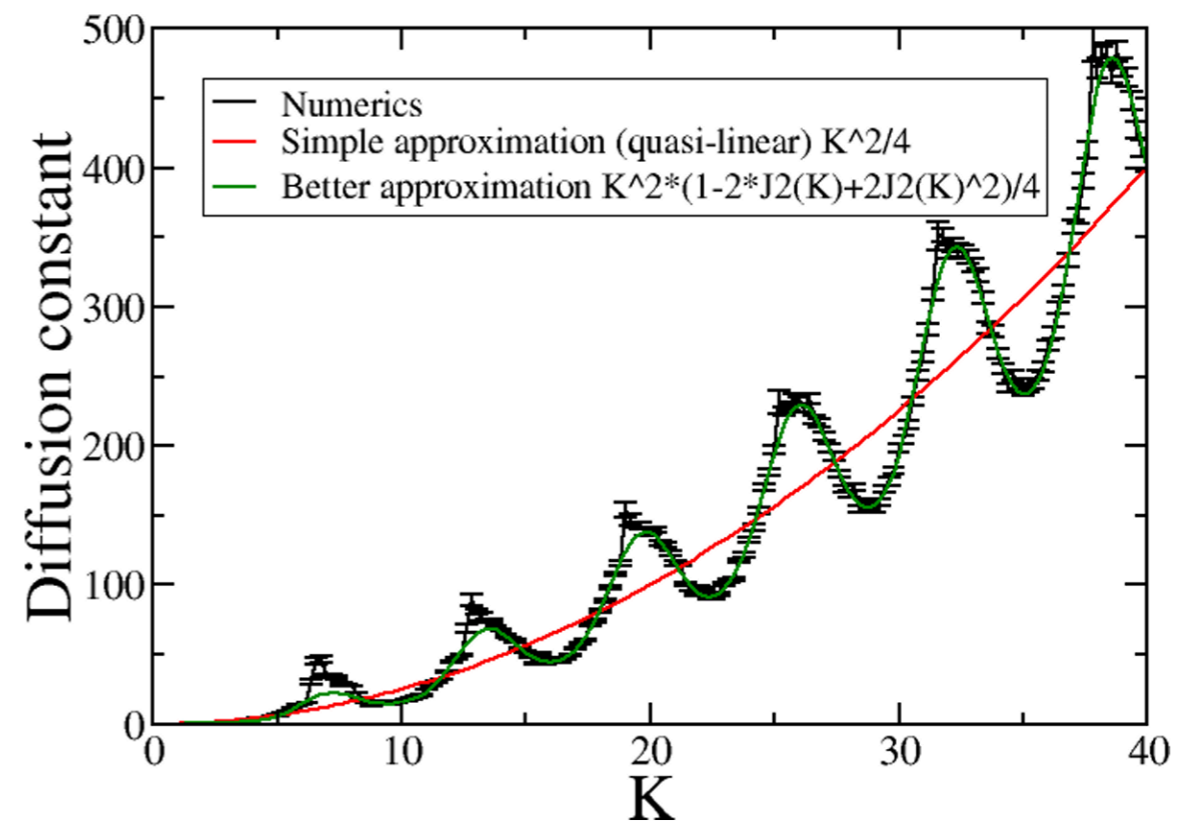


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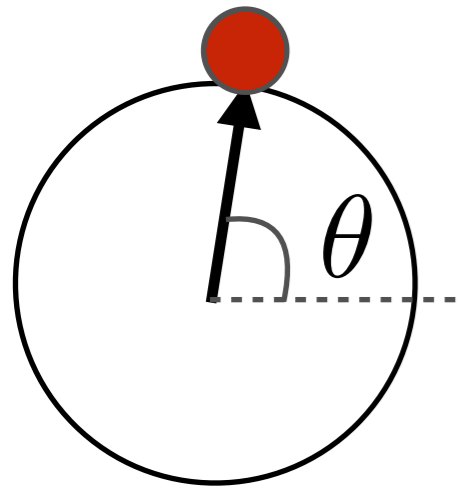
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CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

Kicked rotor



The root-mean-square of a diffusion process

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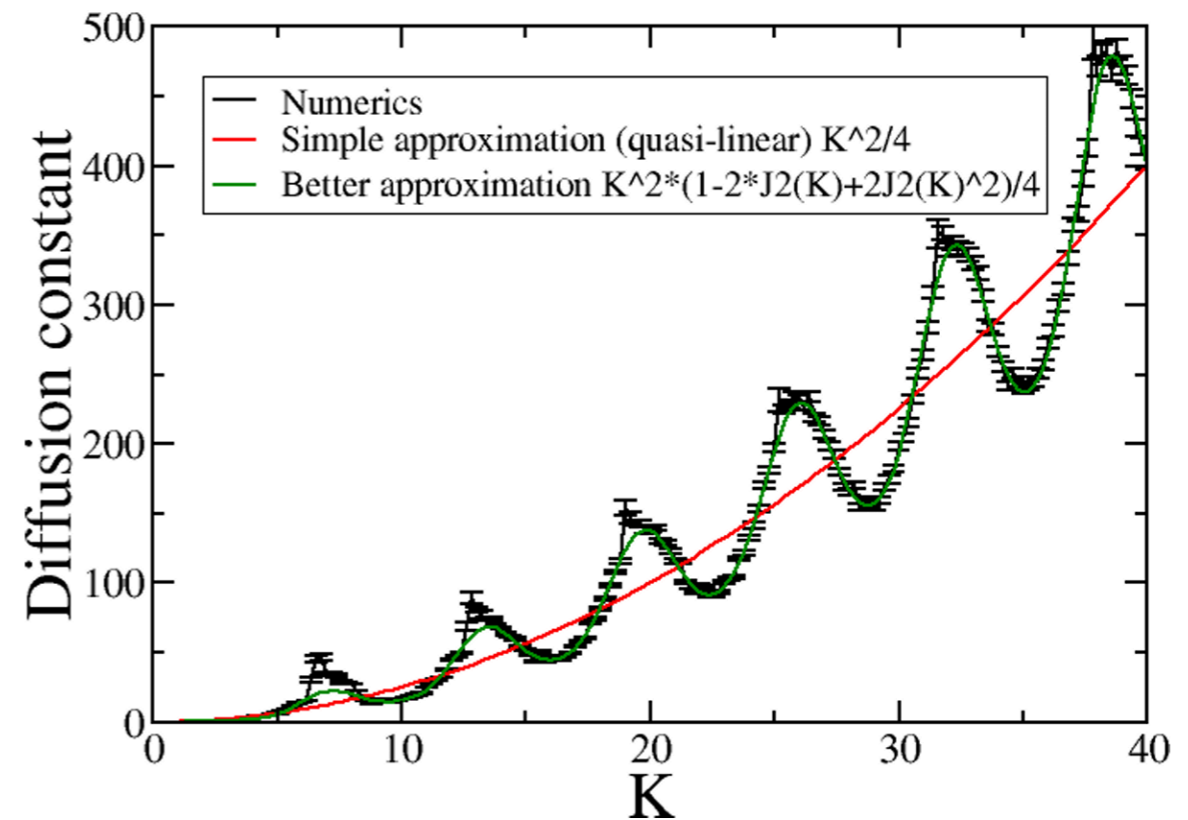
Corrections in the diffusion constant
(Correlations between kicks)

$$D \approx \frac{k^2}{4} (1 - 2J_2(k) + 2J_2^2(k))$$

Rechester and White, PRL (1980)

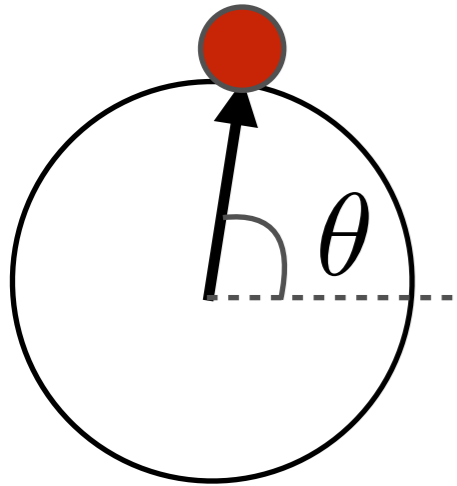
Diffusion in the kicked rotor

$$\langle (p_m - p_0)^2 \rangle_{\text{rms}} = 2Dm$$



CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

Kicked rotor



Lyapunov exponent

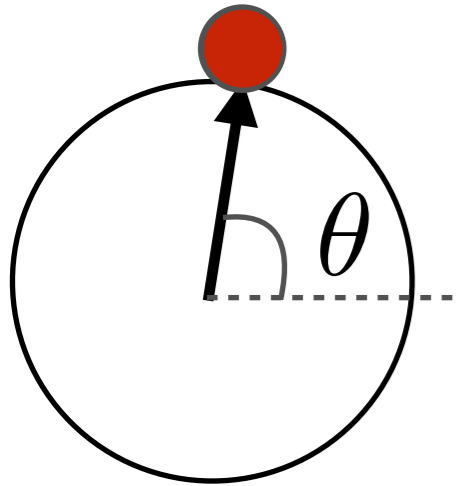
$$d(t) \approx d(0)e^{\lambda(x,p)t},$$

$$d(t) = \sqrt{[x'(t) - x(t)]^2 + [p'(t) - p(t)]^2}$$

$$\lambda = \langle\langle \lambda(x, p) \rangle\rangle = \left\langle\left\langle \lim_{t \rightarrow \infty} \lim_{d(0) \rightarrow 0} \frac{1}{t} \ln \frac{d(t)}{d(0)} \right\rangle\right\rangle$$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

Kicked rotor



Lyapunov exponent

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2 nearby trajectories

$$\begin{cases} p_{n+1} = p_n + K \sin x_n \\ x_{n+1} \equiv x_n + p_{n+1} \pmod{2\pi} \end{cases}$$

$$\eta_n = p'_n - p_n$$

$$\eta_{n+1} = \eta_n + K(\sin x'_n - \sin x_n)$$

$$\begin{cases} p'_{n+1} = p'_n + K \sin x'_n \\ x'_{n+1} \equiv x'_n + p'_{n+1} \pmod{2\pi} \end{cases}$$

$$\xi_n = x'_n - x_n$$

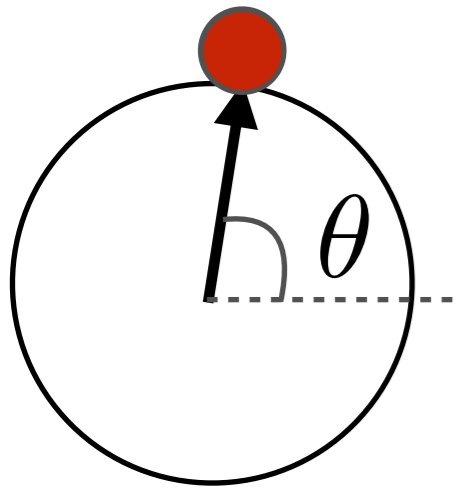
$$\xi_{n+1} = \xi_n + \eta_{n+1}$$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

Kicked rotor

Lyapunov exponent

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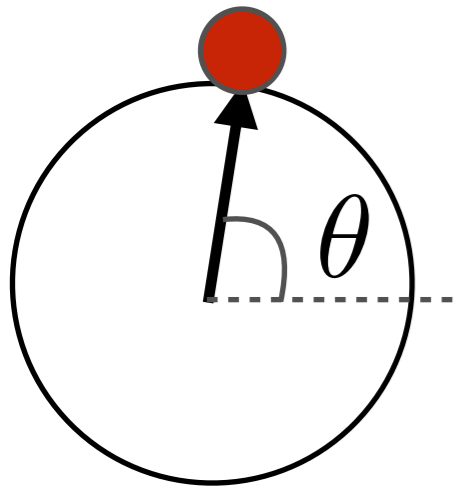
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CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

Kicked rotor

Lyapunov exponent

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$$\eta_n = p'_n - p_n$$

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$$\xi_n = x'_n - x_n$$

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$$\begin{cases} p_{n+1} = p_n + K \sin x_n \\ x_{n+1} \equiv x_n + p_{n+1} \pmod{2\pi} \end{cases}$$

$$\begin{cases} p'_{n+1} = p'_n + K \sin x'_n \\ x'_{n+1} \equiv x'_n + p'_{n+1} \pmod{2\pi} \end{cases}$$

Trig identity

$$\begin{aligned} \sin x'_n - \sin x_n &= \sin(x_n + \xi_n) - \sin x_n \\ &= \sin x_n (\cos \xi_n - 1) + \sin \xi_n \cos x_n. \end{aligned}$$

To leading order
in the deviations

$$\eta_{n+1} = \eta_n + (K \cos x_n) \xi_n$$

$$\xi_{n+1} = \xi_n + \eta_{n+1}$$

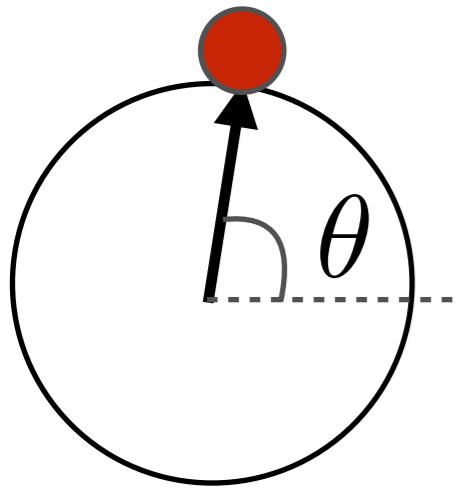
CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

Kicked rotor

Lyapunov exponent

$$\eta_{n+1} = \eta_n + (K \cos x_n) \xi_n$$

$$\xi_{n+1} = \xi_n + \eta_n + (K \cos x_n) \xi_n$$



To leading order
in the deviations

$$\begin{pmatrix} \xi_{n+1} \\ \eta_{n+1} \end{pmatrix} = \begin{pmatrix} 1 & k \cos(x_n) \\ 1 & k \cos(x_n) + 1 \end{pmatrix} \begin{pmatrix} \xi_n \\ \eta_n \end{pmatrix}$$

Eigenvalues $\lambda_{\pm} = 1 + \frac{k \cos(x_n)}{2} \pm \sqrt{k \cos(x_n) \left(1 + \frac{k \cos(x_n)}{4}\right)}$

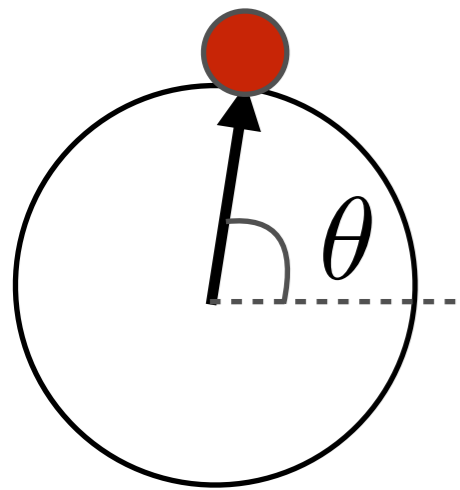
Focusing on $k > 0$

$$\lambda \approx \langle \langle l_+ [k \cos(x_n)] \rangle \rangle = \frac{1}{2\pi} \int_{-\pi}^{\pi} dx \ln |l_+ [k \cos(x)]|$$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED ROTOR

Kicked rotor

Lyapunov exponent



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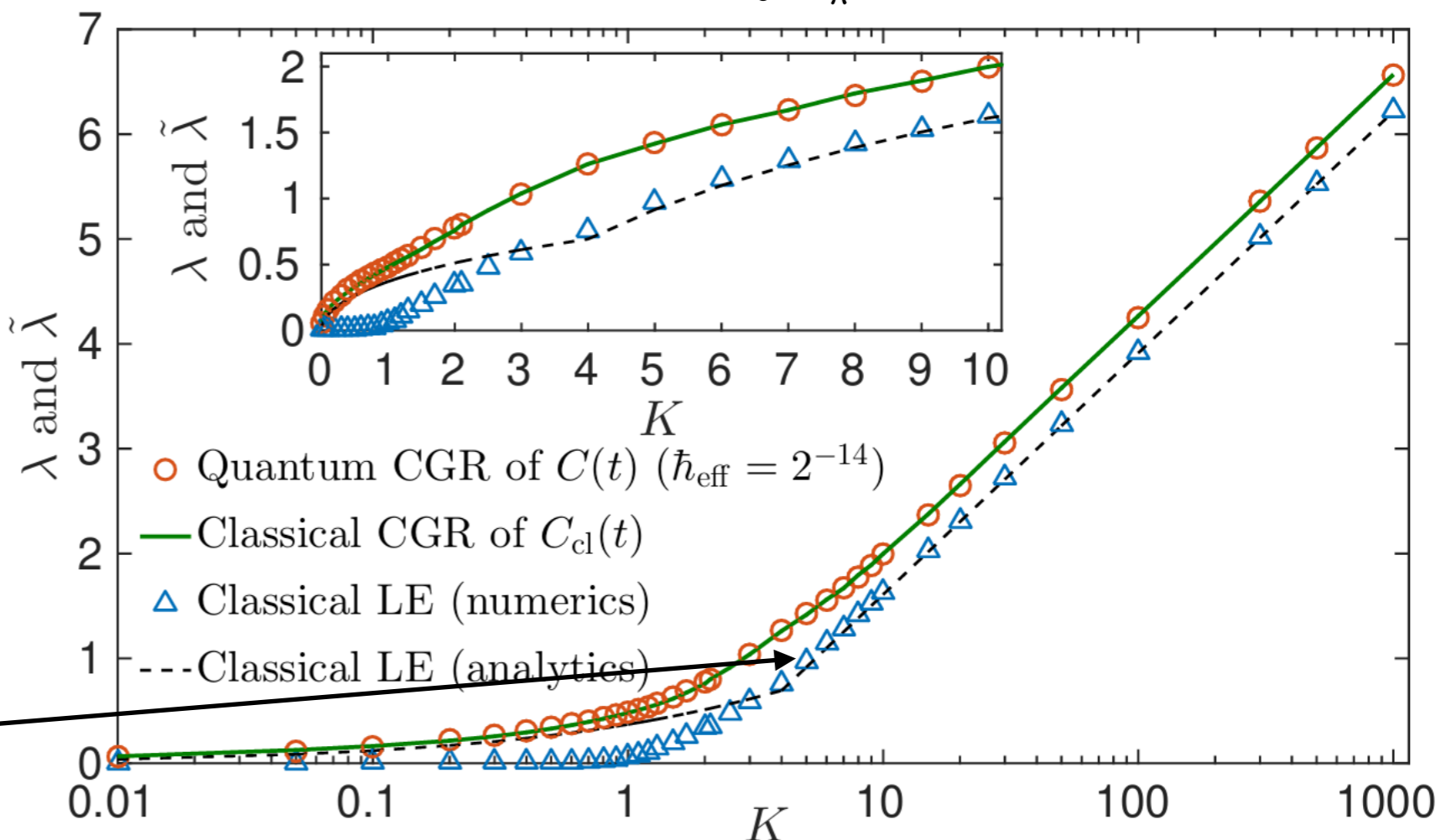
$k > 0$ $\lambda \approx \langle \langle l_+[k \cos(x_n)] \rangle \rangle = \frac{1}{2\pi} \int_{-\pi}^{\pi} dx \ln |l_+[k \cos(x)]|$

At large k

$$\lambda \approx \frac{1}{2\pi} \int_{-\pi}^{\pi} dx \ln(k |\cos(x)|)$$

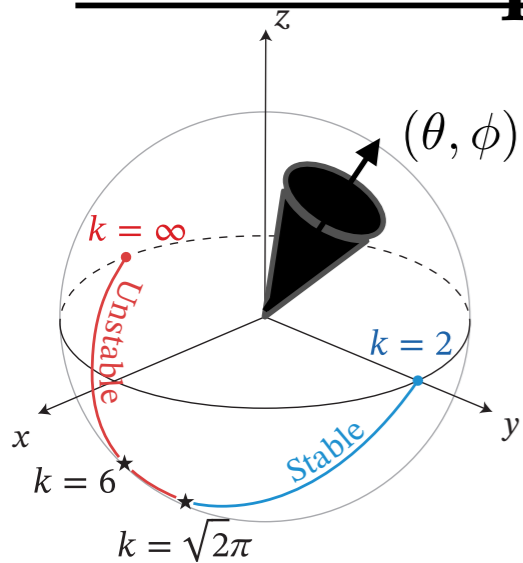
$$= \log(k/2)$$

Focus on **blue data** and
on **black dashed line**



CHAOTIC HAMILTONIAN SYSTEMS: KICKED TOP

Kicked top

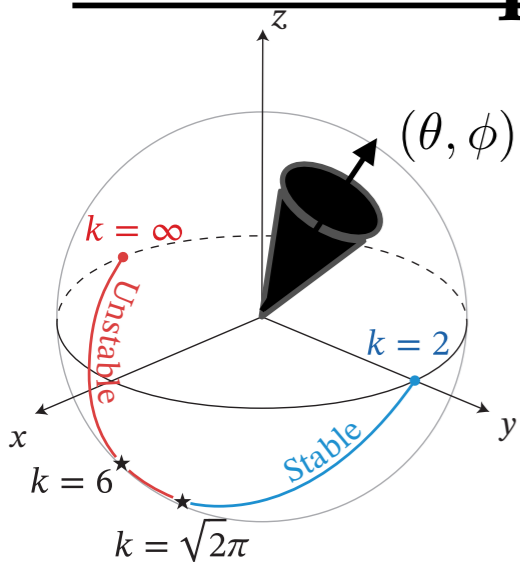


$$H_{\text{KT}} = \alpha J_y + \frac{k J_z^2}{2S} \sum_{n=-\infty}^{\infty} \delta(t - nT)$$

$\alpha = \pi/(2T)$.

CHAOTIC HAMILTONIAN SYSTEMS: KICKED TOP

Kicked top



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$$\{J_i, J_j\} = i\epsilon_{ijk} J_k$$

Conserved angular momentum

$$\{\mathbf{J}^2, H\} = 0$$

Equation of Motion

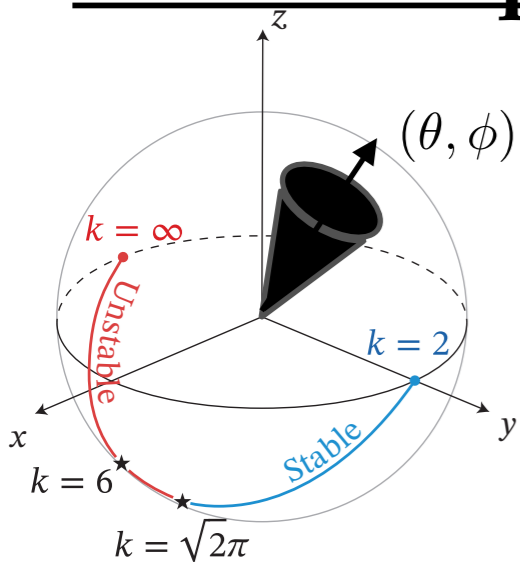
$$\frac{d\mathbf{J}}{dt} = \{\mathbf{J}, H\} = -\mathbf{J} \times \frac{\partial H}{\partial \mathbf{J}}$$

Classical limit $S \rightarrow \infty$

$$x = J_x/S, y = J_y/S, z = J_z/S$$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED TOP

Kicked top



$$H_{\text{KT}} = \alpha J_y + \frac{k J_z^2}{2S} \sum_{n=-\infty}^{\infty} \delta(t - nT)$$

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Integrating the equations, across one period

$$\int_{mT}^{(m+1)T} dt(\dots)$$

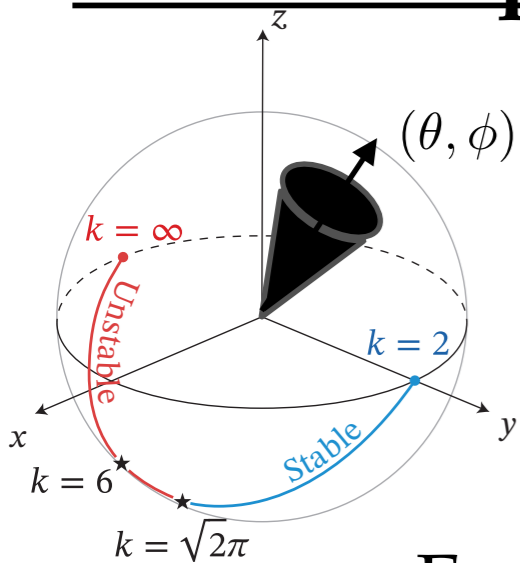
$$x_{n+1} = z_n \cos(kx_n) + y_n \sin(kx_n)$$

$$y_{n+1} = -z_n \sin(kx_n) + y_n \cos(kx_n)$$

$$z_{n+1} = -x_n$$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED TOP

Kicked top



Equation of Motion

$$H_{\text{KT}} = \alpha J_y + \frac{k J_z^2}{2S} \sum_{n=-\infty}^{\infty} \delta(t - nT)$$

$$\alpha = \pi/(2T)$$

$$S \rightarrow \infty \quad x = J_x/S, y = J_y/S, z = J_z/S$$

$$x_{n+1} = z_n \cos(kx_n) + y_n \sin(kx_n)$$

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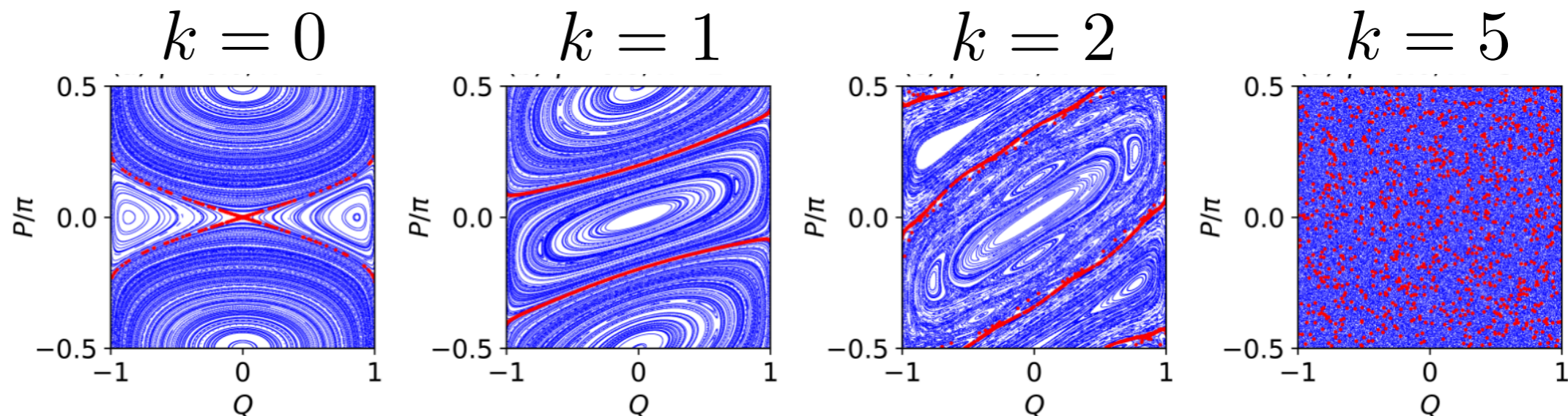
$$z_{n+1} = -x_n$$

Phase space coordinates

$$z = Q$$

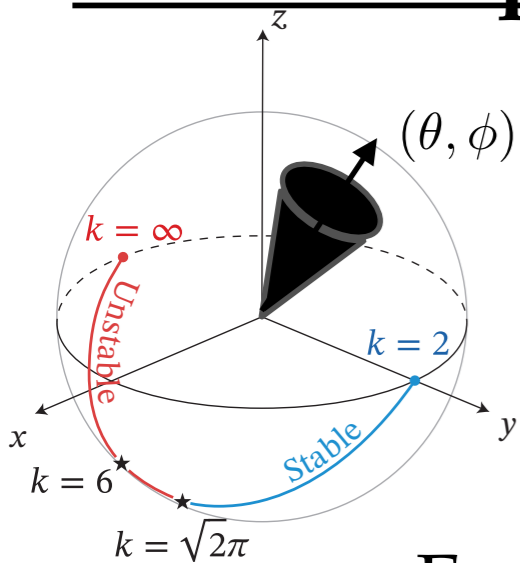
$$x = \sqrt{1 - Q^2} \cos(2P)$$

$$y = \sqrt{1 - Q^2} \sin(2P)$$



CHAOTIC HAMILTONIAN SYSTEMS: KICKED TOP

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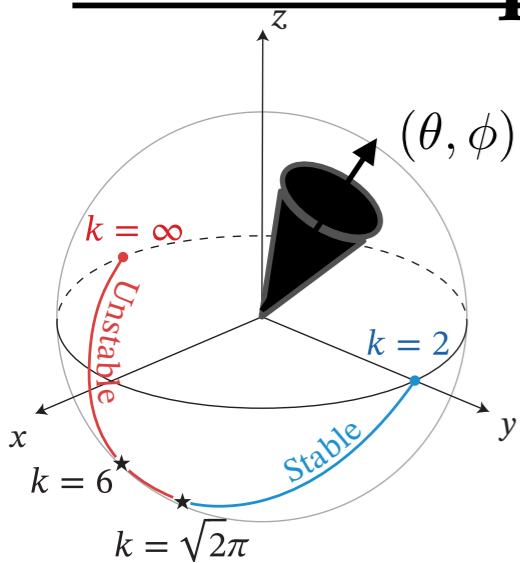
$$z_{n+1} = -x_n$$

Phase space is compact, NO notion of diffusion in the long time limit

But in the limit of short times and large kicking strengths it is diffusive

CHAOTIC HAMILTONIAN SYSTEMS: KICKED TOP

Kicked top



$$H_{\text{KT}} = \alpha J_y + \frac{k J_z^2}{2S} \sum_{n=-\infty}^{\infty} \delta(t - nT)$$

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Diffusion in the kicked top, in the limit of large kicking strength ($k \gg 1$) and SHORT times!

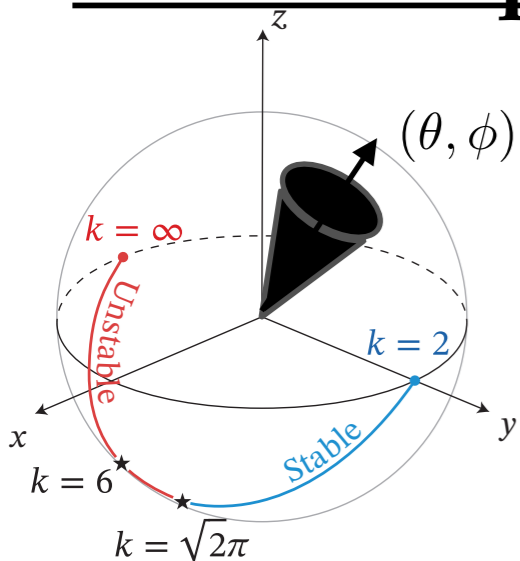
$$\langle [z_{n+1} - z_n]^2 \rangle = \langle [x_n + z_n]^2 \rangle$$

$k \gg 1$, all kicks are random and independent

$$\approx \langle x_n^2 \rangle + \langle z_n^2 \rangle = 2/3$$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED TOP

Kicked top



$$H_{\text{KT}} = \alpha J_y + \frac{k J_z^2}{2S} \sum_{n=-\infty}^{\infty} \delta(t - nT)$$

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$$S \rightarrow \infty \quad x = J_x/S, y = J_y/S, z = J_z/S$$

Diffusion in the kicked top, in the limit of large kicking strength ($k \gg 1$) and SHORT times!

$$\langle [z_{n+1} - z_n]^2 \rangle \approx \langle x(t)^2 \rangle + \langle z(t)^2 \rangle = 2/3$$

$k \gg 1$, all kicks are random and independent

$$\langle [z_{n+1} - z_n]^2 \rangle = 2D$$

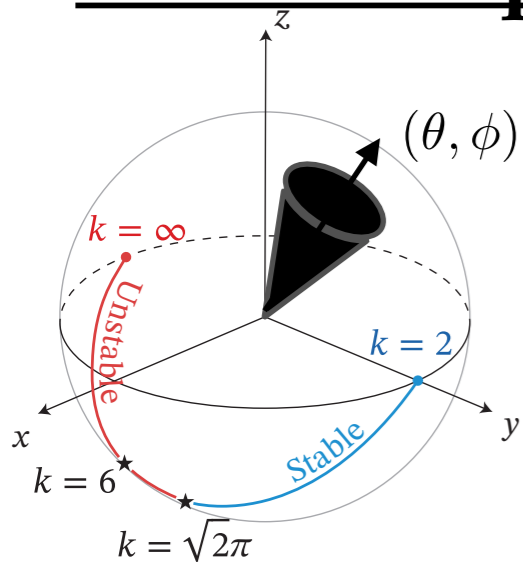
$$D \approx 1/3$$

k independent!

Consequence of compact phase space

CHAOTIC HAMILTONIAN SYSTEMS: KICKED TOP

Kicked top



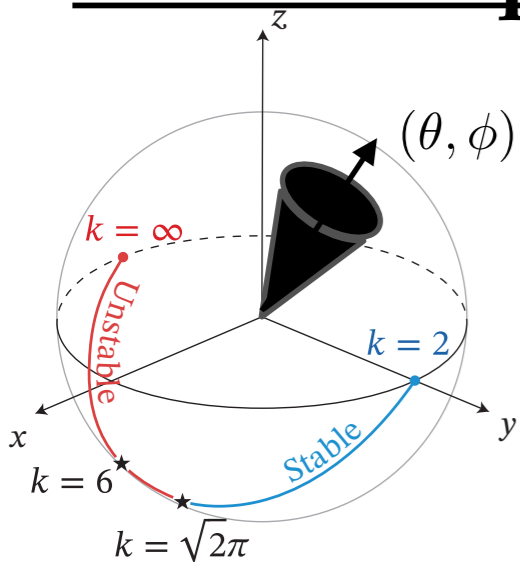
$$H_{\text{KT}} = \alpha J_y + \frac{k J_z^2}{2S} \sum_{n=-\infty}^{\infty} \delta(t - nT)$$

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Stable fixed points of the map $x_{n+1} = x_n = x_0$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED TOP

Kicked top



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Stable fixed points of the map $x_{n+1} = x_n = x_0$

$$z_0 = -x_0 \quad y_0 = x_0 \sin(kx_0) + y_0 \cos(kx_0)$$

$$y_0/x_0 = \sin(kx_0)/(1 - \cos(kx_0)) = \cot(kx_0/2)$$

$$x_0^2 + y_0^2 + z_0^2 = 1$$

$$x_0^2 = \frac{\sin^2\left(\frac{kx_0}{2}\right)}{1 + \sin^2\left(\frac{kx_0}{2}\right)}$$

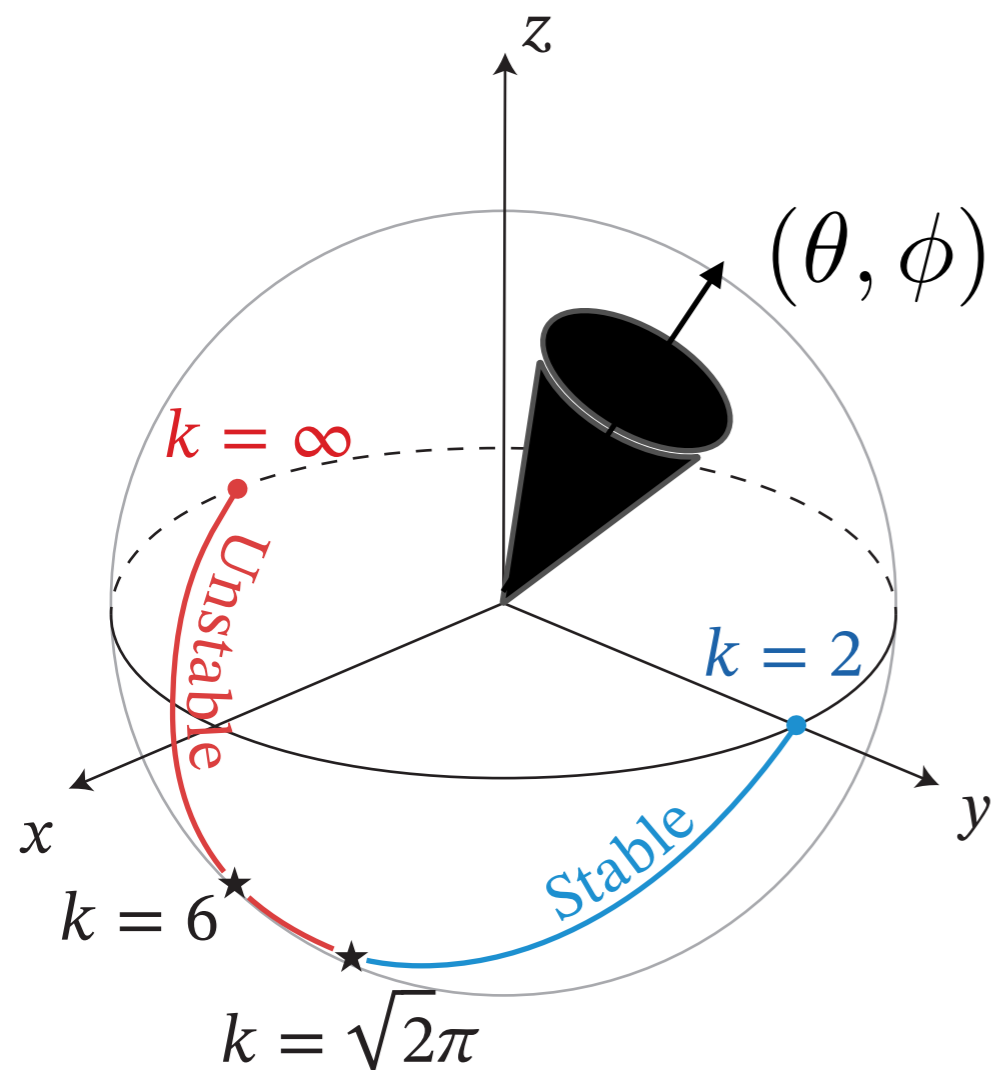
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Stable fixed points of the map



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This must satisfy $\leq 1/2$

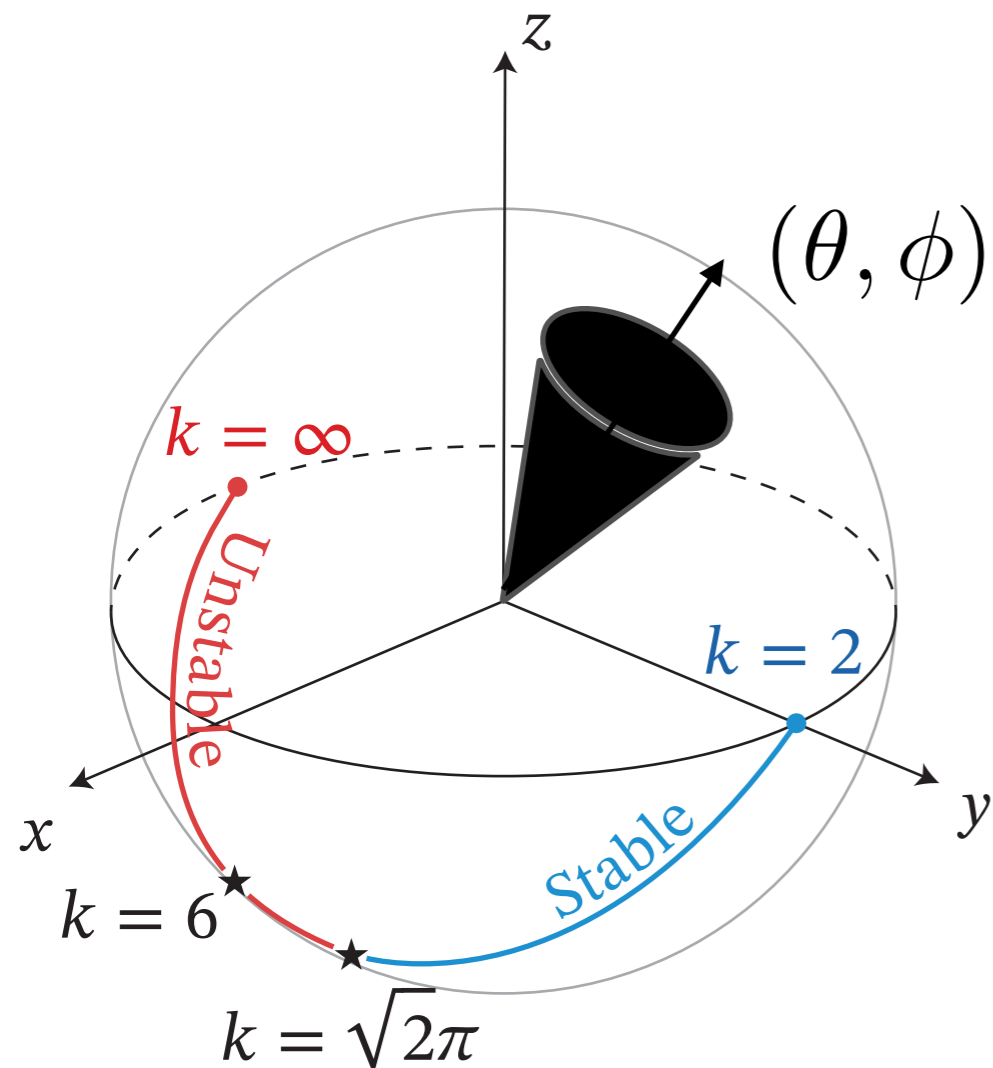
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For at least one solution to exist in the small x_0 limit

$$x_0^2 = (kx_0/2)^2 (1 - (kx_0/2)^2 + \dots)$$

$$k \geq 2 \quad \text{Guarantees one solution exists with} \quad x_0 > 0$$

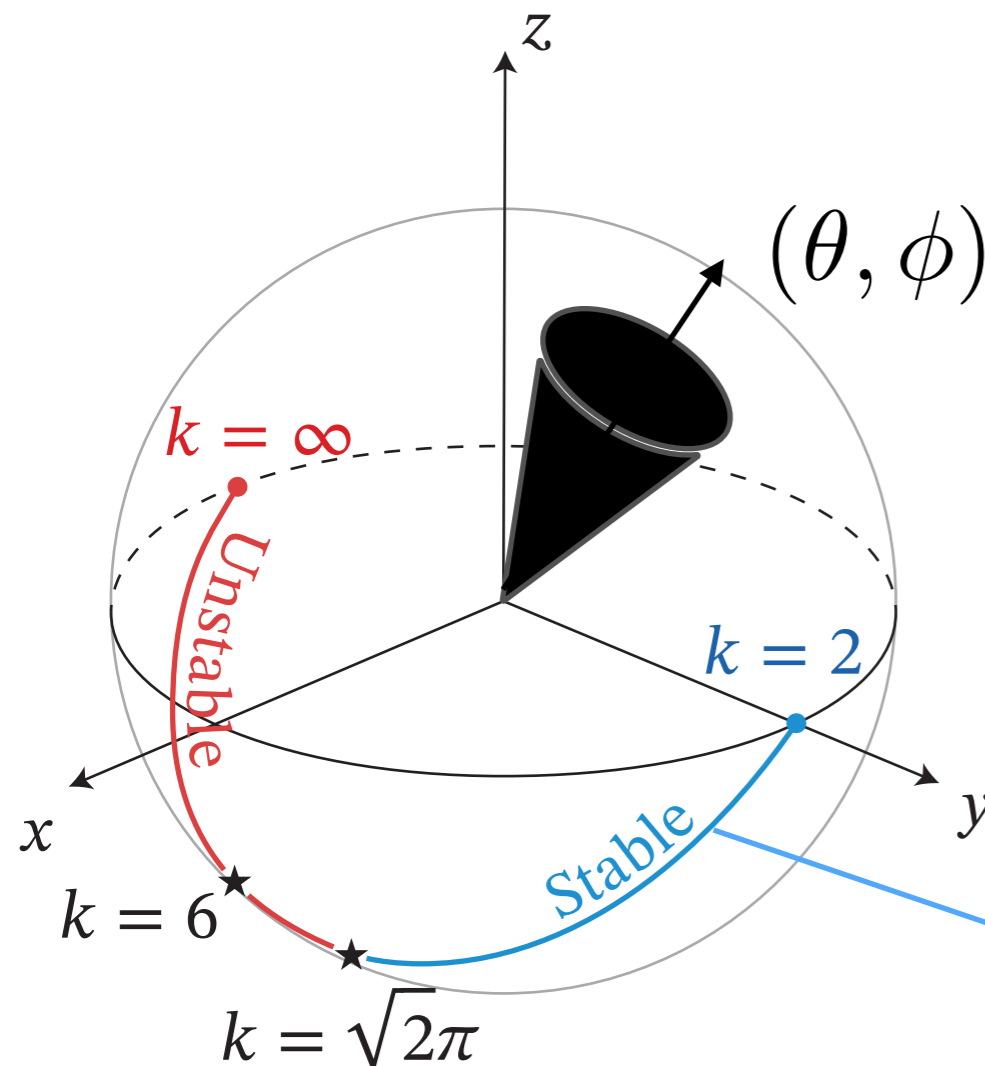
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For at least one solution to exist in the small x_0 limit

$$x_0^2 = (kx_0/2)^2 (1 - (kx_0/2)^2 + \dots)$$

$k \geq 2$ Guarantees one solution exists with $x_0 > 0$

$$x_0 = \frac{\sqrt{3}}{2} \sqrt{k - 2} + O((k - 2)^{3/2})$$

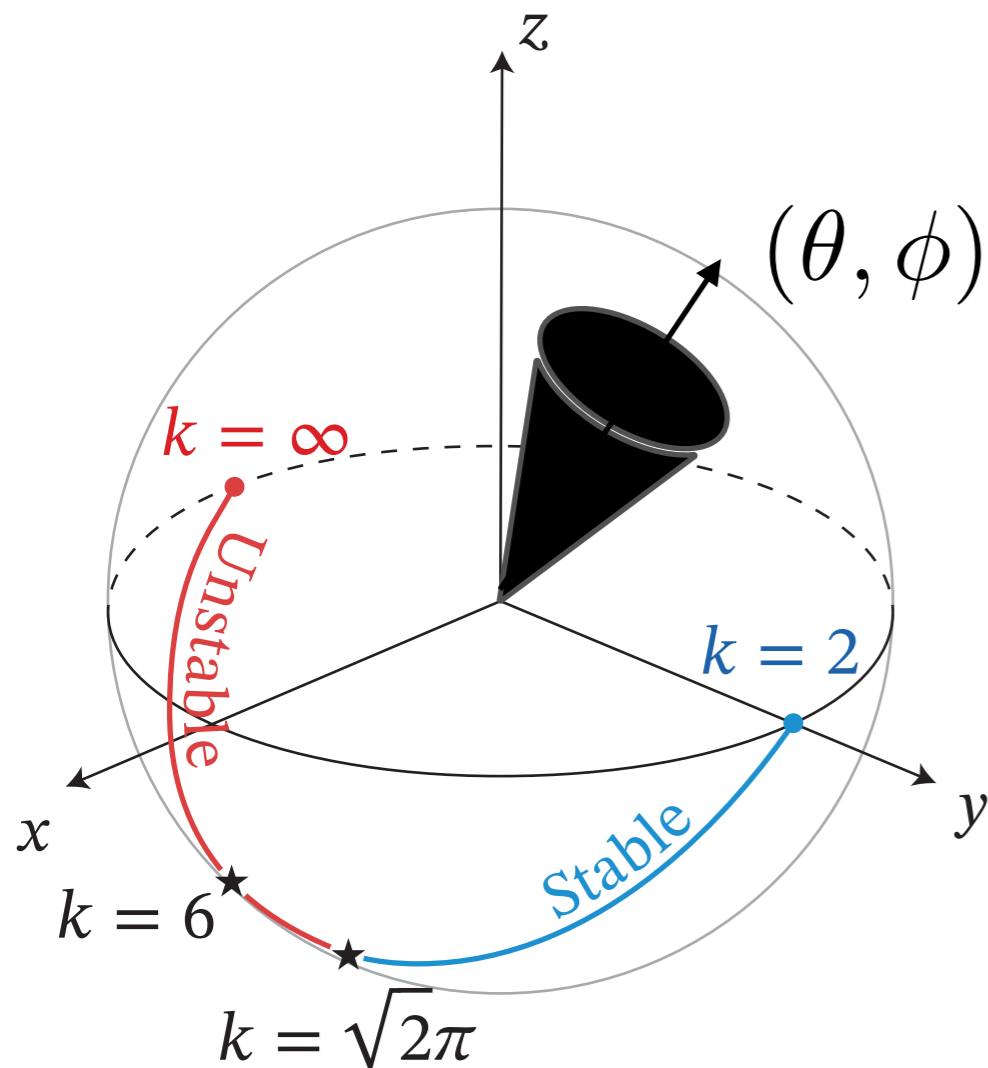
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Unstable fixed points of the map



Linearize about the fixed points

$$x_0^2 = \frac{\sin^2(\frac{kx_0}{2})}{1 + \sin^2(\frac{kx_0}{2})}$$

$$\begin{bmatrix} \delta x' \\ \delta y' \\ \delta z' \end{bmatrix} = \begin{bmatrix} kx_0 \cot(kx_0/2) & \sin(kx_0) & \cos(kx_0) \\ -kx_0 & \cos(kx_0) & -\sin(kx_0) \\ -1 & 0 & 0 \end{bmatrix} \begin{bmatrix} \delta x \\ \delta y \\ \delta z \end{bmatrix}$$

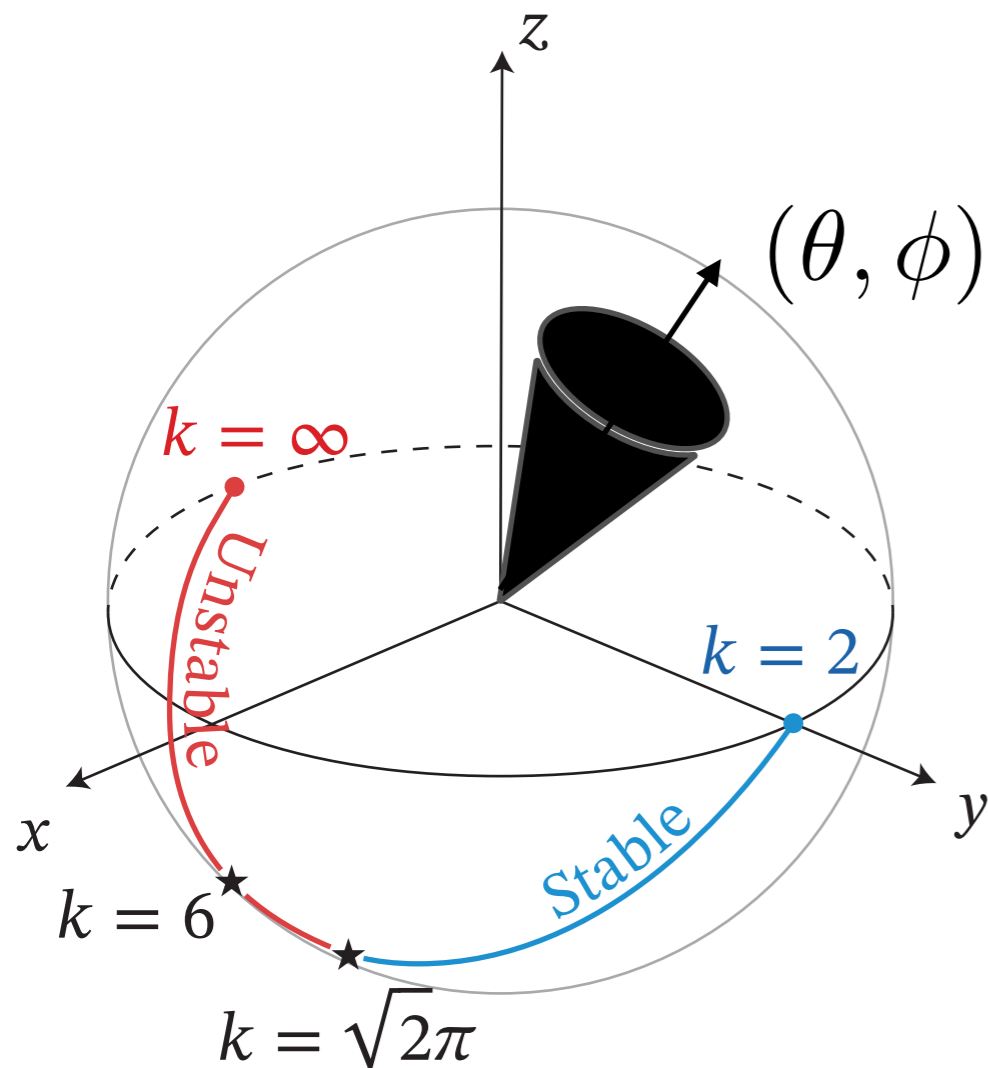
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Unstable fixed points of the map



Linearize about the fixed points $x_0^2 = \frac{\sin^2(\frac{kx_0}{2})}{1 + \sin^2(\frac{kx_0}{2})}$

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Eigenvalues 1 & $\lambda_{\pm}(k) = -\left[h(k) \pm \sqrt{h(k)^2 - 1}\right]$

$$h(k) = \sin^2(kx_0/2) - \frac{kx_0}{2} \cot(kx_0/2)$$

$$|h(k)| < 1 \quad \text{For stability}$$

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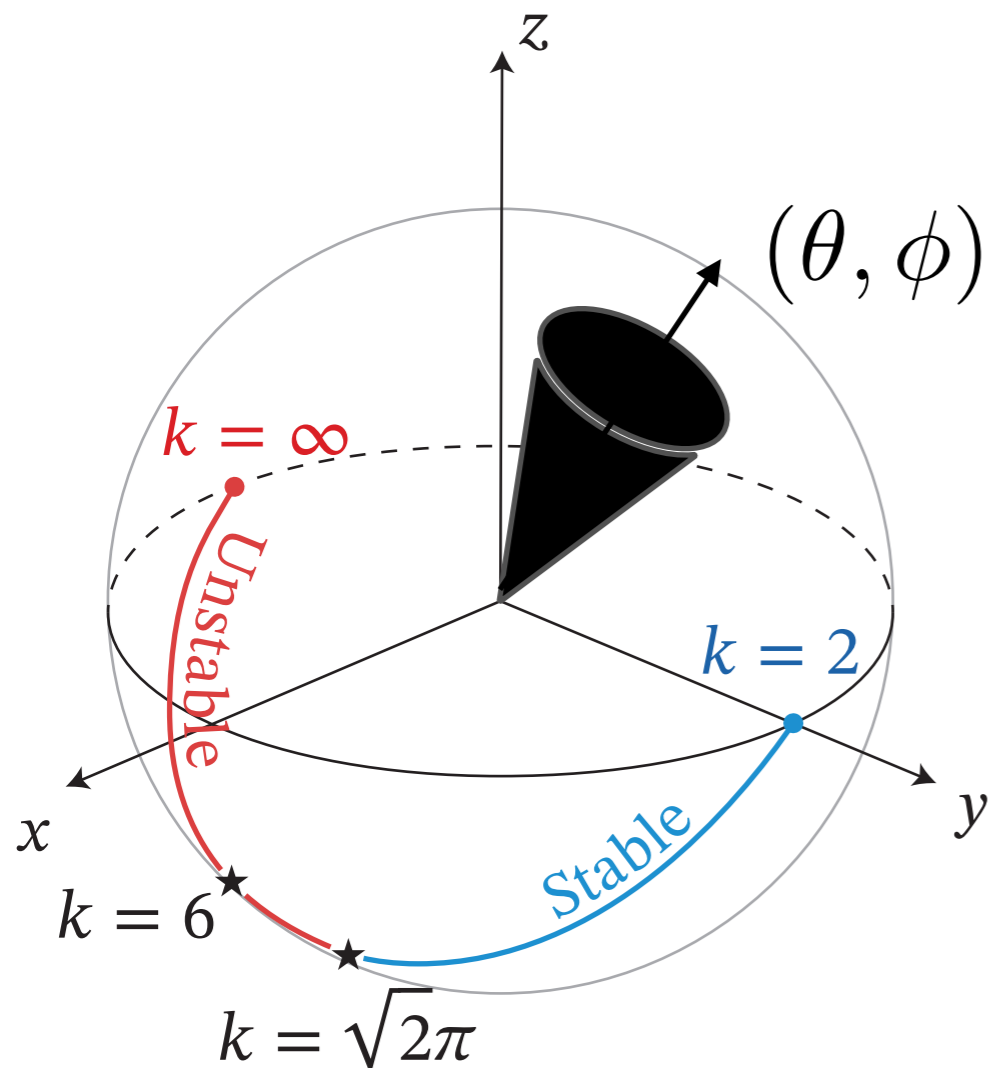
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Becomes unstable when $h(k) = 1$

$$\cos(kx_0/2) = 0$$

$$\sin(kx_0/2) = 1$$



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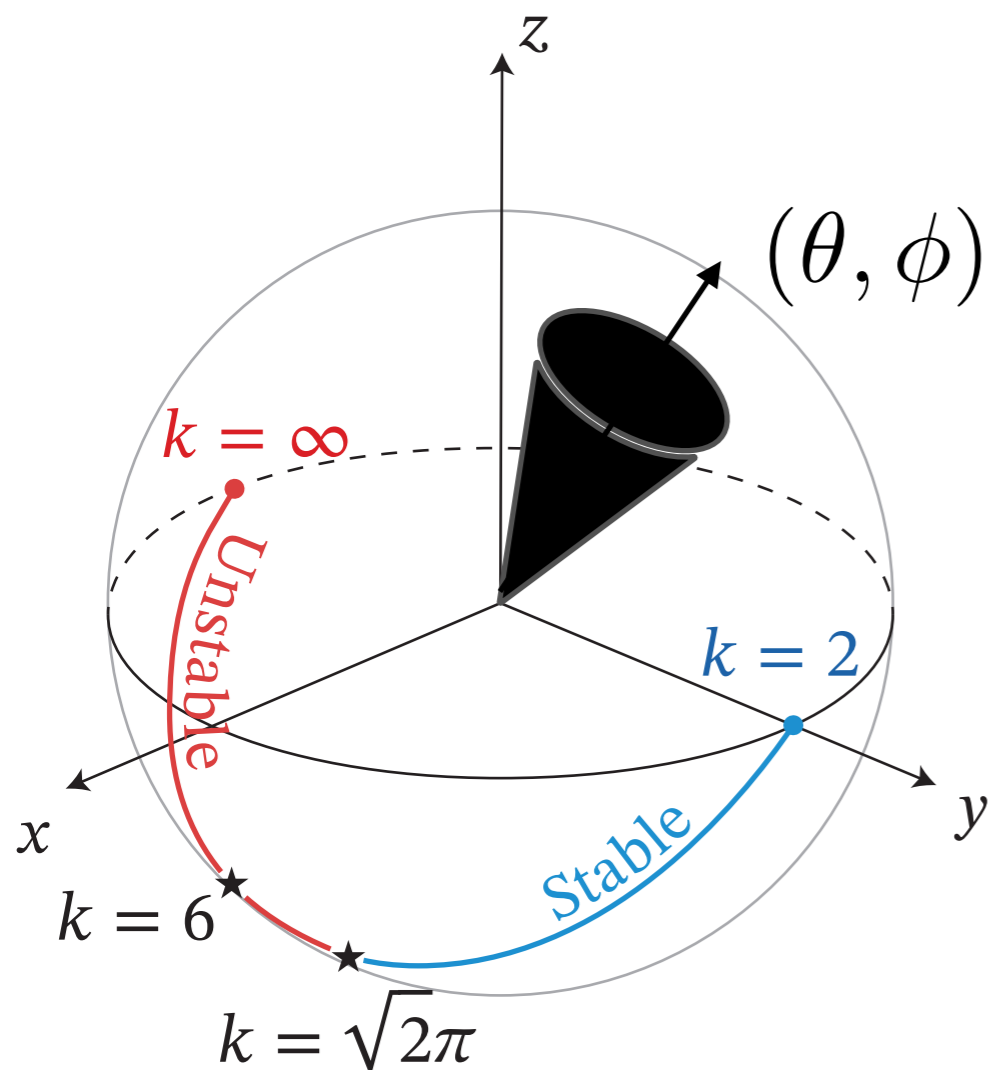
$$\cos(kx_0/2) = 0$$

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$$k_c = \sqrt{2}\pi \approx 4.44$$



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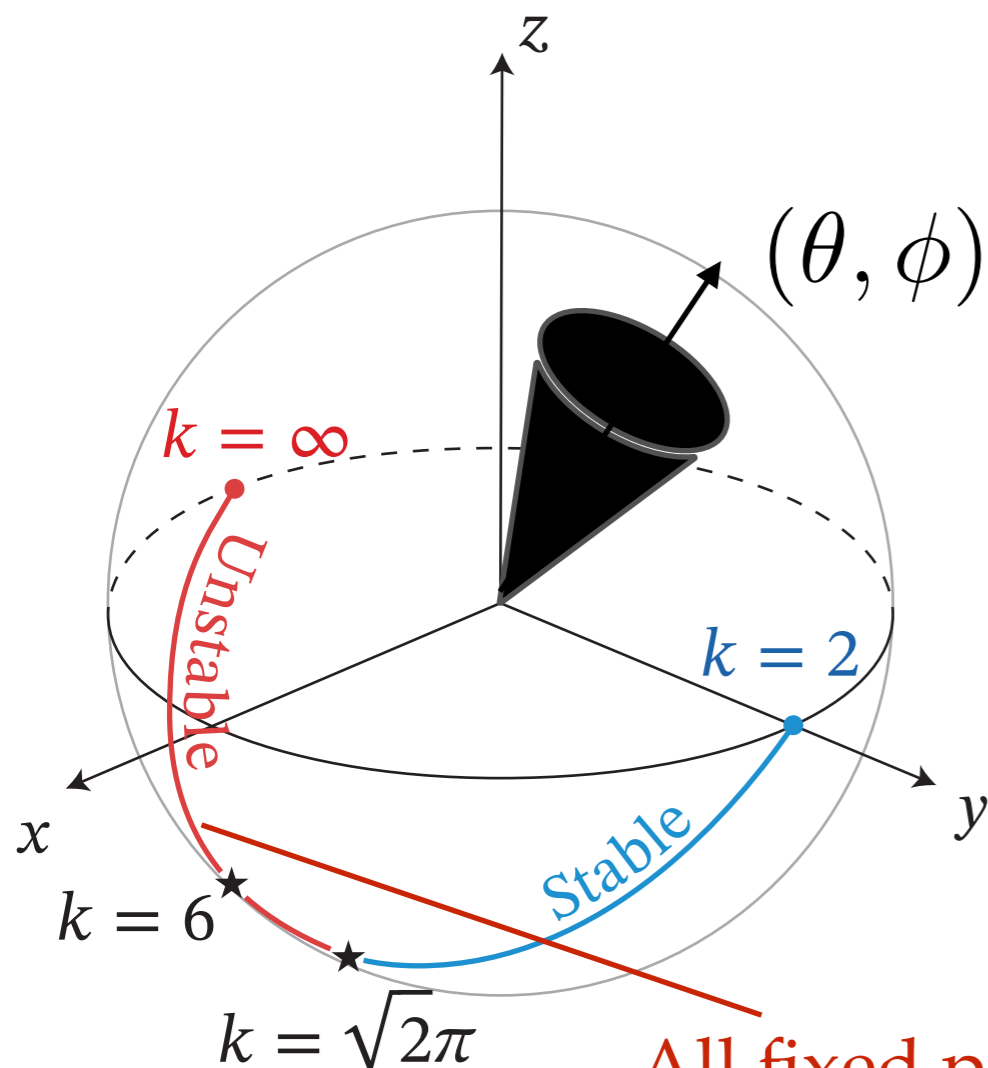
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All fixed points unstable $k > k_c$

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Tangent map

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$$y_{n+1} = -z_n \sin(kx_n) + y_n \cos(kx_n)$$

$$z_{n+1} = -x_n$$

$$\mathbf{r}_n = (x_n, y_n, z_n)$$

$$\mathbf{r}_{n+1} = f(\mathbf{r}_n)\mathbf{r}_n$$

$$\delta \mathbf{r}_n = \mathbf{r}_n - \mathbf{r}'_n$$

$$\delta \mathbf{r}_{n+1} = f'(\mathbf{r}_n)\delta \mathbf{r}_n$$

$$f'(\mathbf{J}_n) = \frac{\partial \mathbf{r}_{n+1}}{\partial \mathbf{r}_n} \quad M_{ij}^n = \frac{\partial r_{n+1}^i}{\partial r_n^j}$$

$$\delta \mathbf{r}_{n+1} = \prod_{j=1}^n f'(\mathbf{r}_j)\delta \mathbf{r}_0$$

CHAOTIC HAMILTONIAN SYSTEMS: KICKED TOP

Kicked top

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Lyapunov exponent

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$$f'(\mathbf{r}_n) = \begin{pmatrix} -kz_n \sin(kx_n) + ky_n \cos(kx_n) & \sin(kx_n) & \cos(kx_n) \\ -kz_n \cos(kx_n) - ky_n \sin(kx_n) & \cos(kx_n) & -\sin(kx_n) \\ -1 & 0 & 0 \end{pmatrix}$$

Define the largest exponent of the product $j_+(n)$

$$\lambda = \log \left(\lim_{n \rightarrow \infty} |j_+(n)|^{1/n} \right)$$

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Largest eigenvalue of \mathbf{f}'

Define the largest exponent of the product $j_+(n)$

$$\lambda = \log\left(\lim_{n \rightarrow \infty} |j_+(n)|^{1/n}\right) \approx \lim_{n \rightarrow \infty} \frac{1}{n} \sum_n \log(|j_+(n)^{\max}|)$$


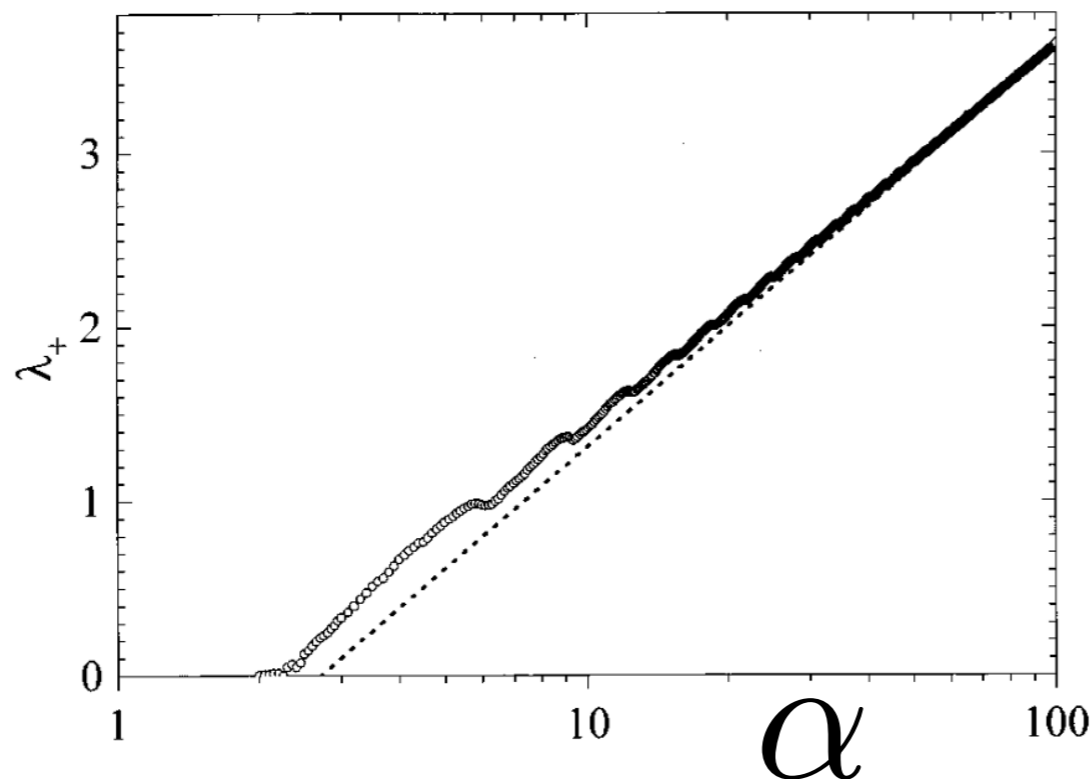
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Large α limit

$$\lambda \approx \log(\alpha \sin(k/2))$$

CHAOTIC HAMILTONIAN SYSTEMS: HEISENBERG SPIN CHAIN

So far only considered single particle problems

Now moving to many-body chaos

CHAOTIC HAMILTONIAN SYSTEMS: HEISENBERG SPIN CHAIN

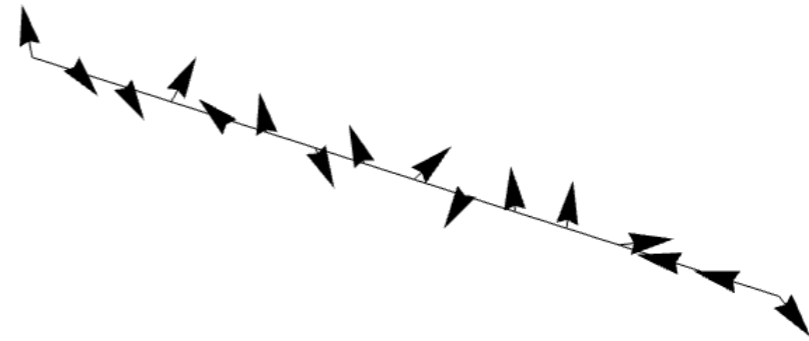
So far only considered single particle problems

Now moving to many-body chaos

Interacting spin chain

$$H = -J \sum_i \mathbf{S}_i \cdot \mathbf{S}_{i+1}$$

$$\frac{d\mathbf{S}_j}{dt} = \{\mathbf{S}_j, H\} = -J(\mathbf{S}_{j-1} + \mathbf{S}_{j+1}) \times \mathbf{S}_j$$



CHAOTIC HAMILTONIAN SYSTEMS: HEISENBERG SPIN CHAIN

So far only considered single particle problems

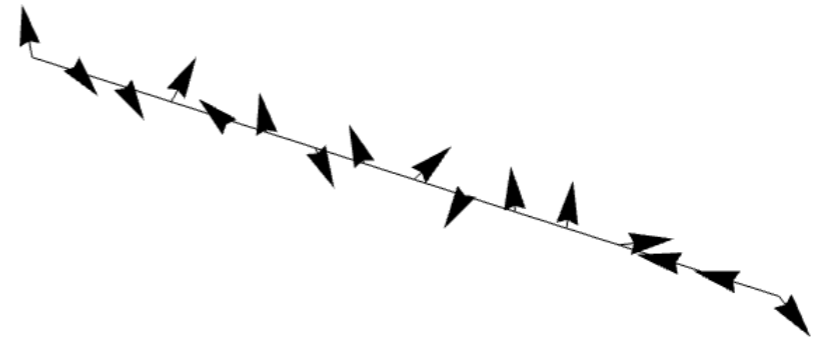
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Most initial states are chaotic $\lambda_{\max} \approx \frac{1}{4} \sqrt{2(J_x^2 + J_y^2 + J_z^2)}$



CHAOTIC HAMILTONIAN SYSTEMS: SPIN MODELS

Now many-body chaos in higher dimensions

Interacting spins models

$$H = \sum_{\langle i,j \rangle} (J_x S_i^x S_j^x + J_y S_i^y S_j^y + J_z S_i^z S_j^z)$$

$\langle i,j \rangle$ Sum over
nearest neighbors

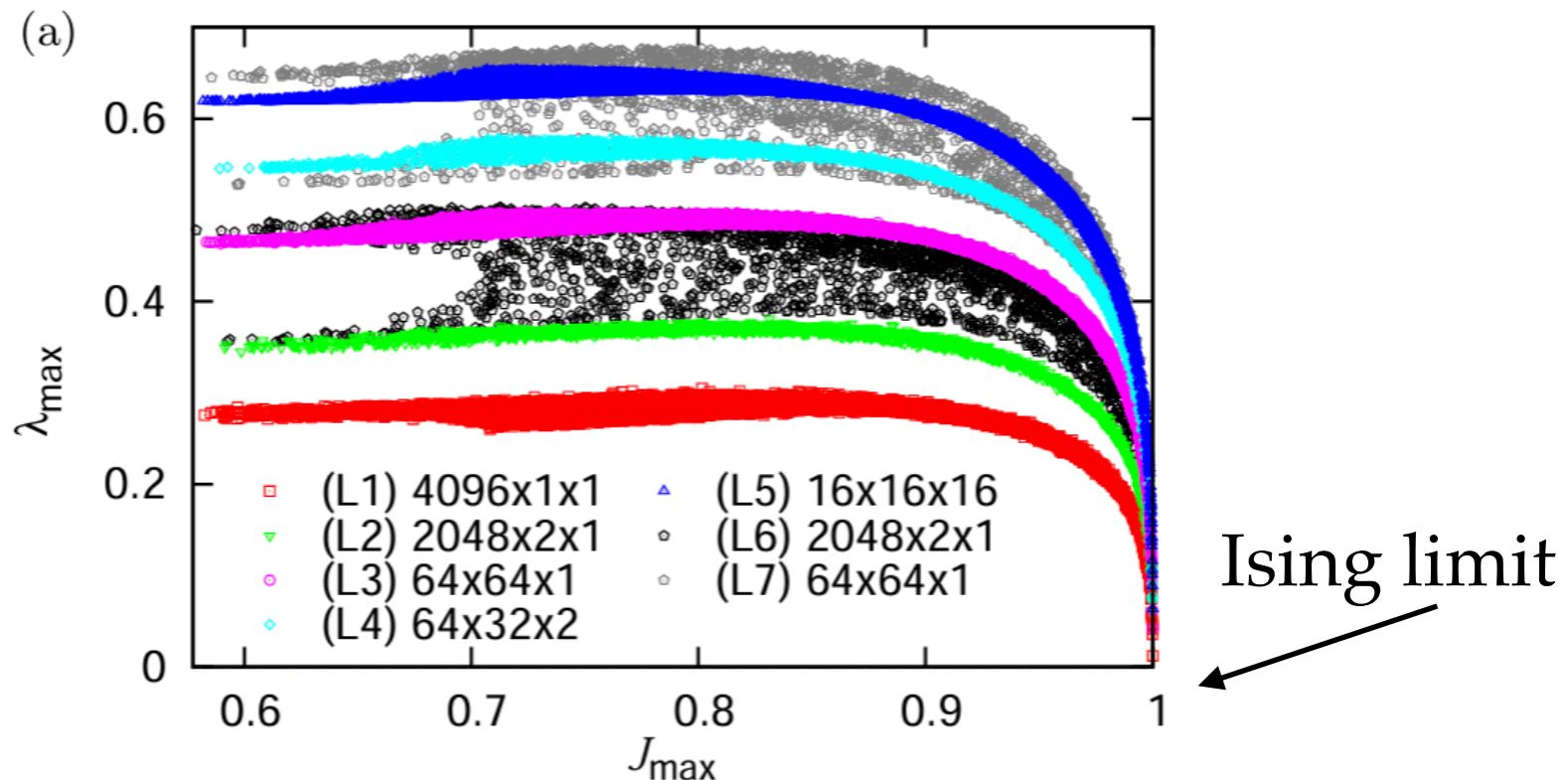
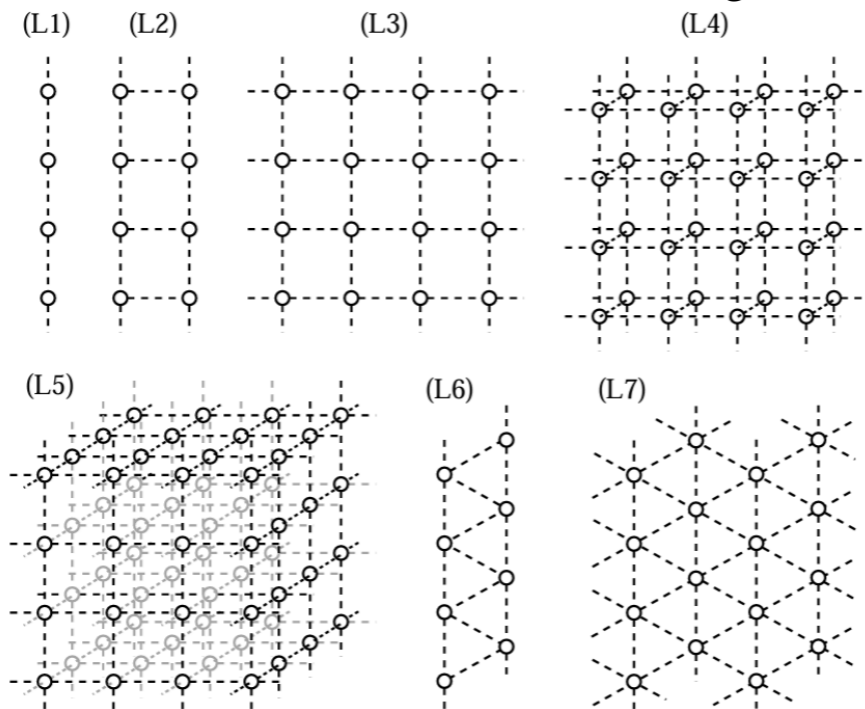
CHAOTIC HAMILTONIAN SYSTEMS: SPIN MODELS

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$\langle i,j \rangle$ Sum over nearest neighbors



$$J_{\max} \equiv \max(|J_x|, |J_y|, |J_z|)$$

Wijn, Hess, Fine, PRL (2012)

Das, Rao, Ramaswamy, EPL (2002)

NUMERICAL EVALUATION OF THE LEADING LYAPUNOV EXPONENT

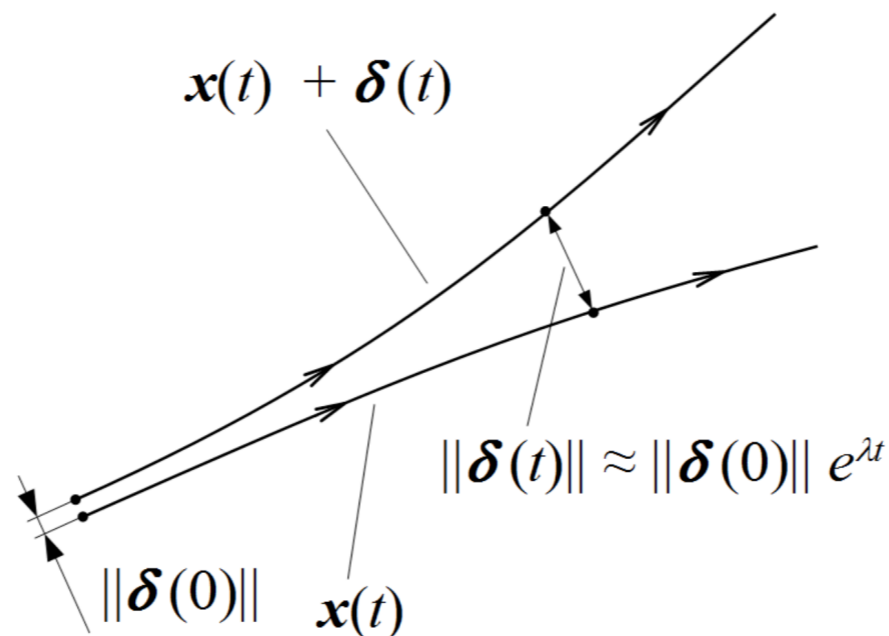
Numerical estimates of Lyapunov exponents can be non-trivial, finite size effects and numerical round off error can be hard to track.

$$\mathbf{x}'(t=0) = \mathbf{x}(0) + \delta \quad \|\mathbf{x}(t) - \mathbf{x}'(t)\| \sim e^{t\lambda_{\text{Lyapunov}}}$$

“Brute Force”

$$\frac{1}{t} \log(\|\mathbf{x}(t) - \mathbf{x}(t')\|) = \lambda_{\text{Lyapunov}} \quad \text{Bad way to go...}$$

Trajectories eventually get too far apart and its not exponential deviation any longer



NUMERICAL EVALUATION OF THE LEADING LYAPUNOV EXPONENT

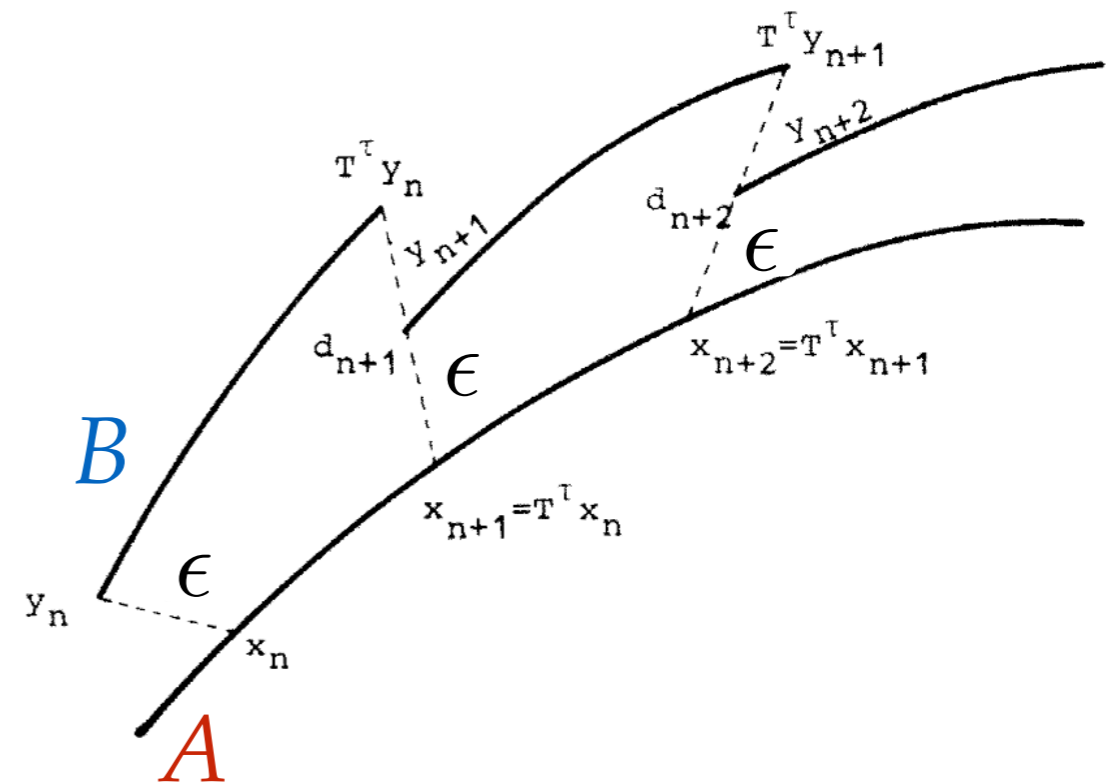
Optimal Numerical Approach: $\lambda = \frac{1}{n\tau} \sum_{j=1}^n \ln \left(\frac{|d_j|}{\epsilon} \right)$

Discretize the time evolution into n, τ steps

Make two copies, A and B . Let A evolve under the time evolution of the problem,

After time τ , reset B to a distance ϵ .

$$d_j = \|x_A(t_j) - x_B(t_j)\|$$



SUMMARY OF WHAT WE HAVE LEARNED

Chaotic dynamics in maps and Hamiltonian dynamics

Single particle models

Logistic map

Bernoulli map

Kicked rotor

Kicked top

Many-body models

Heisenberg spin chain

d-dimensional hyper cubic spin models

OUTLINE

- I. Lecture series layout
- II. Classical chaos
- III. Quantum chaos
- IV. Quantum thermalization
- V. Evading thermalization

CHAOS IN THE QUANTUM LIMIT?

Quantum systems lack the notion of a trajectory, **cannot** define a Lyapunov exponent from the separation between trajectories.

Unitarity enforces time dependent overlaps to be constant, no deviation between wave functions

$$\langle \phi(t) | \psi(t) \rangle = \langle \phi(0) | U_t^\dagger U_t | \psi(0) \rangle = \langle \phi(0) | \psi(0) \rangle$$

Can we define quantum notions of chaos that have a classical chaotic limit?

Are there quantum chaotic systems that don't have a simple classical limit?

CHAOS IN THE QUANTUM LIMIT

Quantum Hamiltonians have an energy spectrum

$$\hat{H} \quad E_i$$

And similarly for Floquet Unitaries, their spectrum is phases

$$\hat{U}_F \quad \lambda_i = e^{i\phi_i}$$

The structure of these matrices and their eigenvalues distinguishes integrable and quantum chaotic systems

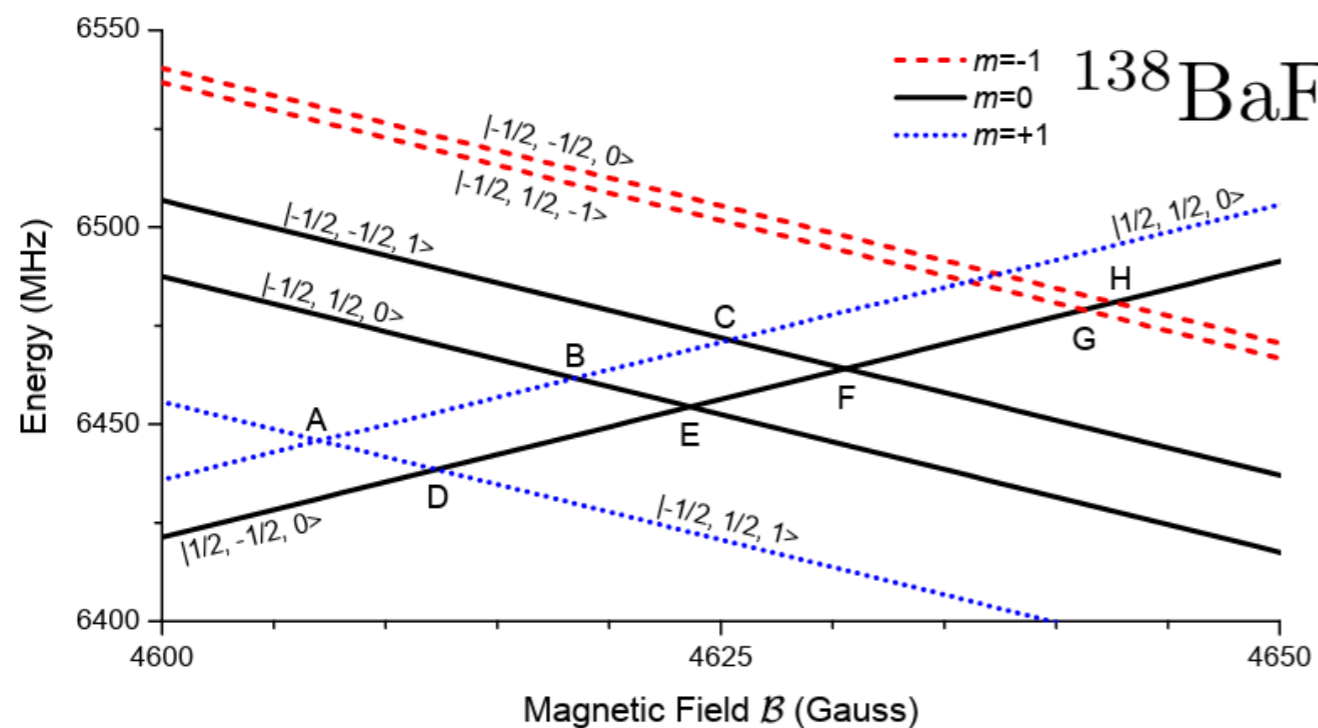
$$\langle i | \hat{H} | j \rangle \quad \langle i | \hat{U}_F | j \rangle$$

ABSENCE OF LEVEL REPULSION: INTEGRABLE SYSTEMS

Examples: $\langle i | \hat{H} | j \rangle = \delta_{ij} E_i = \begin{pmatrix} E_1 & 0 & 0 & \dots \\ 0 & E_2 & 0 & \dots \\ 0 & 0 & E_3 & \dots \\ \vdots & \vdots & \vdots & \ddots \end{pmatrix}$

Integrable Hamiltonians are exactly solvable,
NO off-diagonal elements

Level crossings between Zeeman multiplets



$$E_1, E_2, E_3, \dots$$

Can be become equal as we
vary **one** parameter

ABSENCE OF LEVEL REPULSION: INTEGRABLE SYSTEMS

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Integrable Hamiltonians are exactly solvable,
NO off-diagonal elements

E_1, E_2, E_3, \dots Can be become equal as we
vary **one** parameter

Treat as N independent levels.

Level spacing $s_i = E_i - E_{i-1}$

What is the distribution of the level spacing? $P(s_i)$

ABSENCE OF LEVEL REPULSION: INTEGRABLE SYSTEMS

Treat as N independent levels.

Level spacing $s_i = E_i - E_{i-1}$

What is the distribution of the level spacing? $P(s_i)$

$P(s)$ = Probability of finding the **nearest neighbor of level E** in dE at $E+s$
= $P(\text{we find a level between } E+s \text{ and } E+s+ds \text{ given one at } E)$
* $P(\text{no other level is between } E \text{ and } E+S)$
= $P(s | E) * P(\text{no other level is between } E \text{ and } E+S)$

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= $P(s | E) * P(\text{no other level is between } E \text{ and } E+S)$

Assume a constant density of states

$P(s|E) = \mu(s)$ Independent of E , as the density of states is constant

$\mu(s) = \mu_0$ Assuming all levels are independent

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What is the distribution of the level spacing? $P(s_i)$

$P(s) = P(s | E) * P(\text{no other level is between } E \text{ and } E+S)$

$$P(s) = \mu_0 \left[1 - \int_0^s ds' P(s') \right] = \mu_0 \int_s^\infty ds' P(s')$$

Used the normalization $\int_0^\infty ds P(s) = 1$

Can now solve the differential equation for $P(s)$

ABSENCE OF LEVEL REPULSION: INTEGRABLE SYSTEMS

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What is the distribution of the level spacing? $P(s_i)$

$$P(s) = \mu_0 \left[1 - \int_0^s ds' P(s') \right] = \mu_0 \int_s^\infty ds' P(s')$$

$$\frac{dP}{ds} = \mu_0 (P(\infty) - P(s)) = -\mu_0 P(s)$$

Normalizing and scaling

$$P(s) = A e^{-\mu_0 s}$$

$$\int_0^\infty ds P(s) = 1 \quad \bar{s} = 1$$

$$P(s) = e^{-s}$$

$$P(0) > 0$$

LEVEL REPULSION: HERMITIAN MATRIX

Consider interactions between level 1 and 2.

Start with a 2x2 matrix

$$H_{ij} = \begin{pmatrix} E_1 & V \\ V^* & E_2 \end{pmatrix}$$

Eigenvalues

$$E_{\pm} = \frac{1}{2}(E_1 + E_2) \pm \sqrt{\frac{1}{4}(E_1 - E_2)^2 + V^*V}$$

Level spacing

$$|E_+ - E_-| = 2\sqrt{\frac{1}{4}(E_1 - E_2)^2 + V^*V}$$

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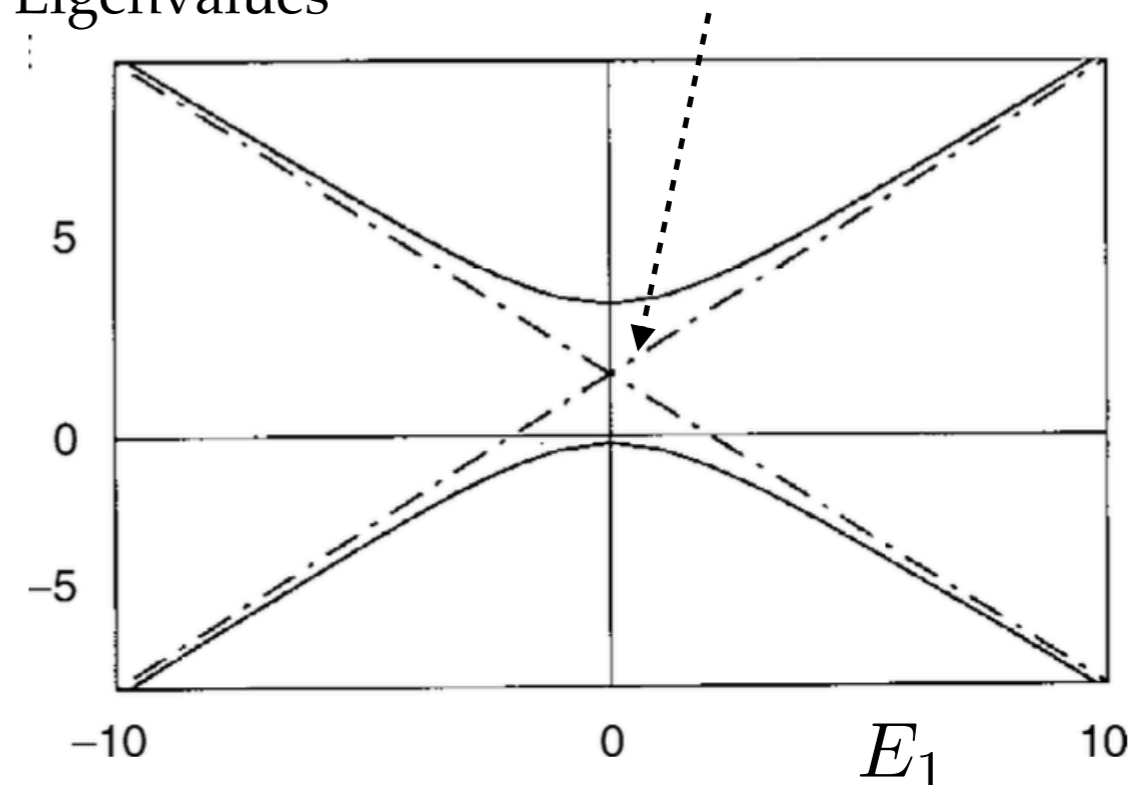
Eigenvalues

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Tuning all three

$$(E_1 - E_2), V^*, V$$

Eigenvalues



Level spacing

$$|E_+ - E_-| = 2\sqrt{\frac{1}{4}(E_1 - E_2)^2 + V^*V}$$

Requires tuning 3 non-negative terms
to make the levels cross $|E_+ - E_-| = 0$

Co-dimension of the avoided crossing $n = 3$

LEVEL REPULSION: REAL MATRIX

Now make $V \in \mathbb{R}$ $H_{ij} = \begin{pmatrix} E_1 & V \\ V & E_2 \end{pmatrix}$

Eigenvalues

$$E_{\pm} = \frac{1}{2}(E_1 + E_2) \pm \sqrt{\frac{1}{4}(E_1 - E_2)^2 + V^2}$$

Level spacing

$$|E_+ - E_-| = 2\sqrt{\frac{1}{4}(E_1 - E_2)^2 + V^2}$$

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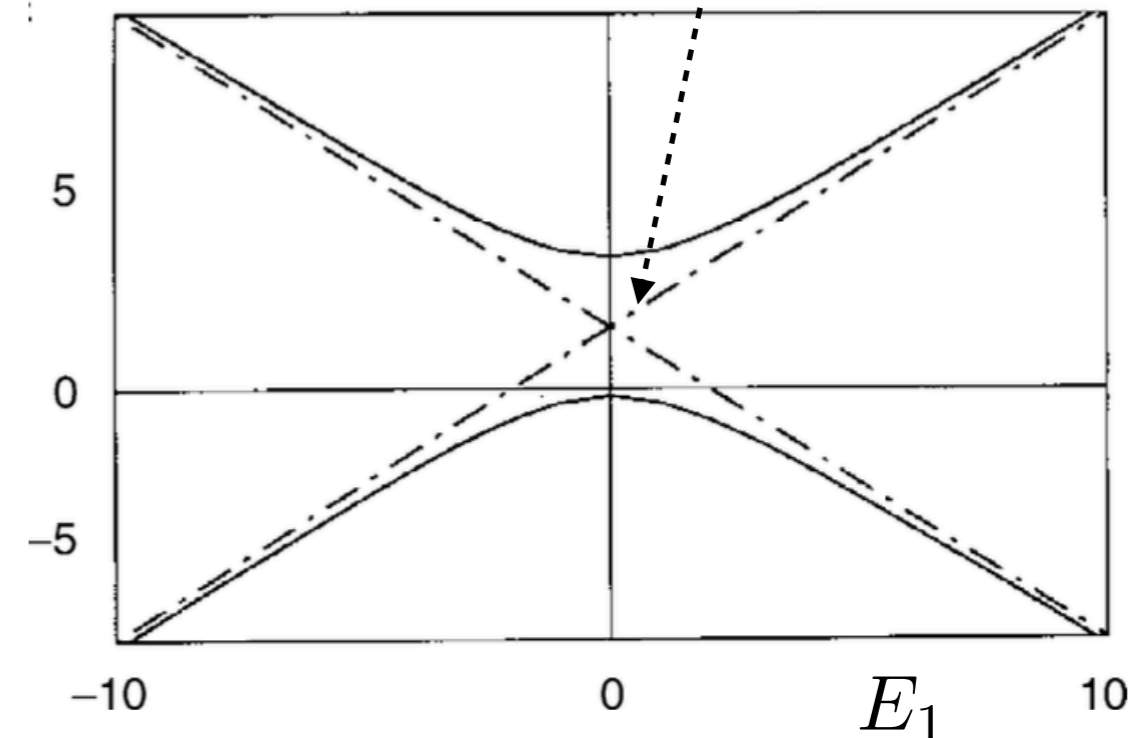
$$|E_+ - E_-| = 2\sqrt{\frac{1}{4}(E_1 - E_2)^2 + V^2}$$

Requires tuning 2 non-negative terms to make the levels cross $|E_+ - E_-| = 0$

Co-dimension of the avoided crossing $n = 2$

Tuning two parameters
 $(E_1 - E_2), V$

Eigenvalues



LEVEL REPULSION, SYMMETRIES, AND RANDOM MATRICES

Thinking more broadly, this crossing will be embedded in a larger matrix. The levels will have contributions from some kind of randomness; e.g. disorder, kicking, measurements

This implies we need to think about:

(1) The symmetries of the Hamiltonian

(2) The statistics of random matrices

$$\begin{pmatrix} E_1 & V & & \\ V^* & E_2 & & \\ 0 & 0 & E_3 & \dots \\ \vdots & \vdots & \vdots & \ddots \\ \vdots & \vdots & \vdots & \ddots \end{pmatrix}$$

SYMMETRY AND LEVEL REPULSION

(1) Consider the symmetry properties of the matrices formed from Hamiltonians matrix elements $H_{ij} = \langle i | \hat{H} | j \rangle$

The 10 fold Cartan classification, includes time reversal, spin-rotation, and particle-hole symmetries.

		Symmetry			Dimension d				
AZ		\mathcal{T}	Ξ	Π	1	2	3	4	Time reversal \mathcal{T}
Unitary (no TR) Hermitian Matrix	A	0	0	0	0	Z	0	Z	Particle-hole Ξ
	AIII	0	0	1	Z	0	Z	0	
	Orthogonal (has TR) Real matrix	AI	1	0	0	0	0	0	
BDI	1	1	1	Z	0	0	0		
D	0	1	0	Z ₂	Z	0	0		
DIII	-1	1	1	Z ₂	Z ₂	Z	0		
Symplectic (has TR)	AII	-1	0	0	0	Z ₂	Z ₂	Z	
	CII	-1	-1	1	Z	0	Z ₂	Z ₂	
	C	0	-1	0	0	Z	0	Z ₂	
	CI	1	-1	1	0	0	Z	0	

RANDOM MATRIX THEORY AND ENSEMBLES

(2) The statistics of random matrices: Overwhelming evidence of universality between classically chaotic systems and their quantum limit having universal level statistics

Universality of level repulsion can be described using random matrix theory

RANDOM MATRIX THEORY AND ENSEMBLES

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Universality of level repulsion can be described using random matrix theory

Interested in the probability distribution of the Hamiltonian's matrix elements

$P(H) = P(E_1)P(E_2)P(V)$ (for real symmetric matrices)

$P(H) = P(E_1)P(E_2)P(V, V^*)$ (for Hermitian matrices)

$$H_{ij} = \begin{pmatrix} E_1 & V \\ V & E_2 \end{pmatrix}$$

$$H_{ij} = \begin{pmatrix} E_1 & V \\ V^* & E_2 \end{pmatrix}$$

In general $P(H) = C e^{-A \text{Tr}(H^2)}$

Our aim is to use this to derive the distribution of level spacings

RANDOM MATRIX THEORY AND ENSEMBLES

$$P(H) = C e^{-A \text{Tr}(H^2)}$$

Our aim is to use this to derive the distribution of level spacings

First relate non-diagonal and diagonal matrices via a unitary transformation

For real symmetric matrices this ensemble will be symmetric under all orthogonal transformations

$$P(H) = P(H'), \quad H' = OHO^{-1}$$

Focus on 2x2 matrices

$$H_{ij} = \begin{pmatrix} E_1 & V \\ V & E_2 \end{pmatrix} \quad E_{\pm} = \frac{1}{2}(E_1 + E_2) \pm \sqrt{\frac{1}{4}(E_1 - E_2)^2 + V^2}$$

H is diagonalized with the orthogonal transformation

$$O = \begin{pmatrix} \cos \Theta & -\sin \Theta \\ \sin \Theta & \cos \Theta \end{pmatrix}$$

$$\begin{aligned} E_1 &= E_+ \cos^2 \Theta + E_- \sin^2 \Theta \\ E_2 &= E_+ \sin^2 \Theta + E_- \cos^2 \Theta \\ V &= (E_+ - E_-) \cos \Theta \sin \Theta \end{aligned}$$

$$P(E_+, E_-) = \det \frac{\partial(E_1, E_2, V)}{\partial(E_+, E_-, \Theta)} P(E_1, E_2, V)$$

RANDOM MATRIX THEORY AND ENSEMBLES

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RANDOM MATRIX THEORY AND ENSEMBLES

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$$E_1, E_2, V^*, V \qquad E_{\pm} = \frac{1}{2}(E_1 + E_2) \pm \sqrt{\frac{1}{4}(E_1 - E_2)^2 + V^*V}$$

For Hermitian matrices this ensemble will be symmetric

under all unitary transformations $P(H) = P(H') \quad H' = U H U^{-1}$

H is diagonalized with the unitary transformation $U = \begin{pmatrix} \cos \Theta & -e^{-i\phi} \sin \Theta \\ e^{i\phi} \sin \Theta & \cos \Theta \end{pmatrix}$

$$E_1 = E_+ \cos^2 \Theta + E_- \sin^2 \Theta$$

$$E_2 = E_+ \sin^2 \Theta + E_- \cos^2 \Theta$$

$$V = (E_+ - E_-) e^{i\phi} \cos \Theta \sin \Theta$$

$$P(E_+, E_-) = \det \frac{\partial(E_1, E_2, V, V^*)}{\partial(E_+, E_-, \Theta, \phi)} P(E_1, E_2, V, V^*)$$

RANDOM MATRIX THEORY AND ENSEMBLES

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RANDOM MATRIX THEORY AND ENSEMBLES

Summary

For real symmetric matrices, Gaussian Orthogonal ensemble (GOE)

$$P(E_+, E_-) = \det \frac{\partial(E_1, E_2, V)}{\partial(E_+, E_-, \Theta)} P(E_1, E_2, V) = C |E_+ - E_-| e^{-A \text{Tr} H^2}$$

For Hermitian matrices, Gaussian Unitary ensemble (GUE)

$$P(E_+, E_-) = \det \frac{\partial(E_1, E_2, V, V^*)}{\partial(E_+, E_-, \Theta, \phi)} P(E_1, E_2, V, V^*) = C |E_+ - E_-|^2 e^{-A \text{Tr} H^2}$$

Power law is set by the **co-dimension**
of the avoided level crossing ($n-1$), general result

$$P(E_+, E_-) = C |E_+ - E_-|^{n-1} e^{-A \text{Tr} H^2}$$

RANDOM MATRIX THEORY AND ENSEMBLES

Focused on 2x2 matrices so far

$$P(E_+, E_-) = C |E_+ - E_-|^{n-1} e^{-A \text{Tr} H^2}$$

Power law is set by the co-dimension
of the avoided level crossing, general result

Generalizing to $N \times N$ matrices

$$P(E_1, E_2, \dots, E_N) = \text{const} \prod_{\mu < \nu}^{1 \dots N} |E_\mu - E_\nu|^{n-1} \exp \left(-A \sum_{\mu=1}^N E_\mu^2 \right)$$

$n = 2$ Orthogonal, GOE

$n = 3$ Unitary, GUE

$n = 5$ Symplectic, GSE

RANDOM MATRIX THEORY AND ENSEMBLES

Focused on 2x2 matrices so far

$$P(E_+, E_-) = C |E_+ - E_-|^{n-1} e^{-A \text{Tr} H^2}$$

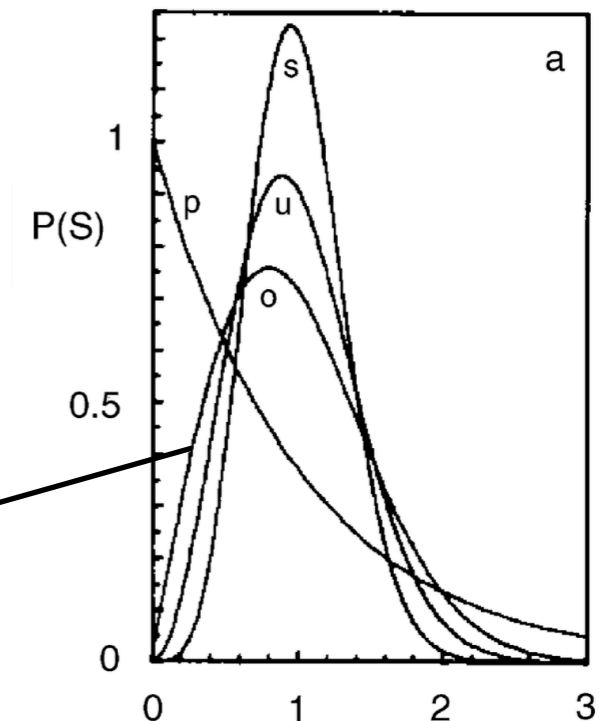
From this, we can compute

$$P(s) = \int_{-\infty}^{\infty} dE_+ \int_{-\infty}^{\infty} dE_- \delta(s - |E_+ - E_-|) P(E_+, E_-)$$

$$= \frac{\pi}{2} s e^{-\frac{\pi s^2}{4}} \quad \text{Orthogonal}$$

$$= \frac{32}{\pi^2} s^2 e^{-\frac{4s^2}{\pi}} \quad \text{Unitary}$$

Level repulsion appears through $P(s=0) = 0$



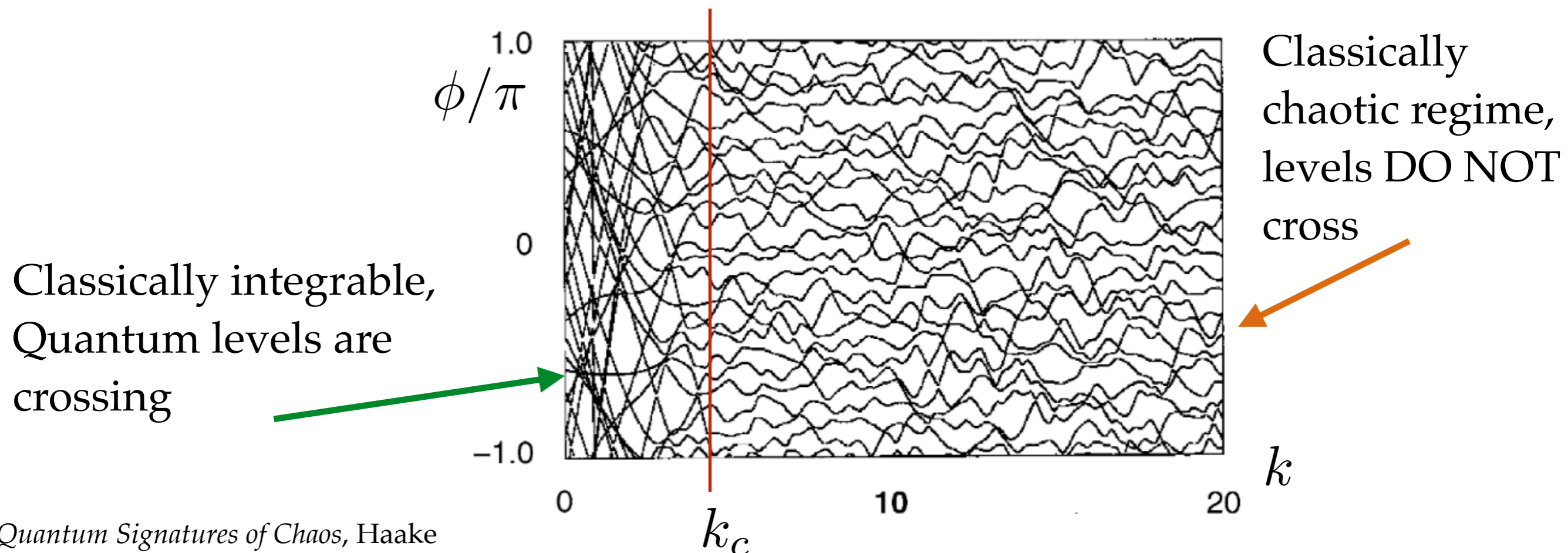
STATISTICS AND LEVEL REPULSION

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Quantum kicked top $\hat{H}_{\text{KT}} = \alpha \hat{J}_y + \frac{k \hat{J}_z^2}{2S} \sum_{n=-\infty}^{\infty} \delta(t - nT)$ $\hat{U}_{\text{KT}} = e^{-i\alpha \hat{J}_y T} e^{-ik \hat{J}_z^2 / (2S)}$

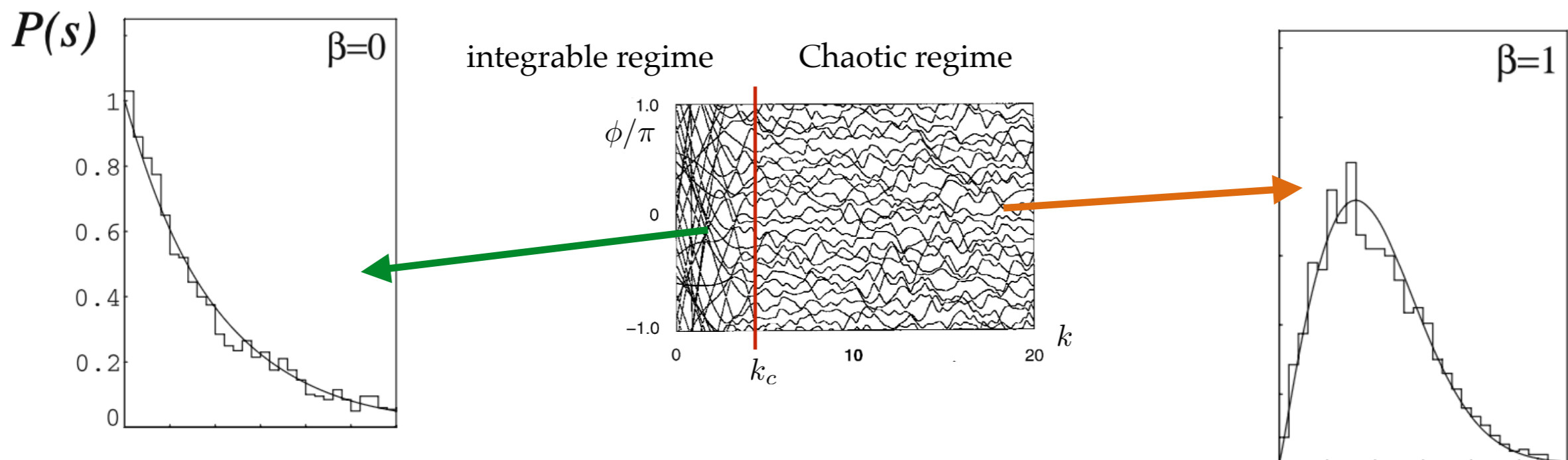


STATISTICS AND LEVEL REPULSION

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Consider the distribution of level spacings $P(s_i)$ $s_i = E_i - E_{i-1}$

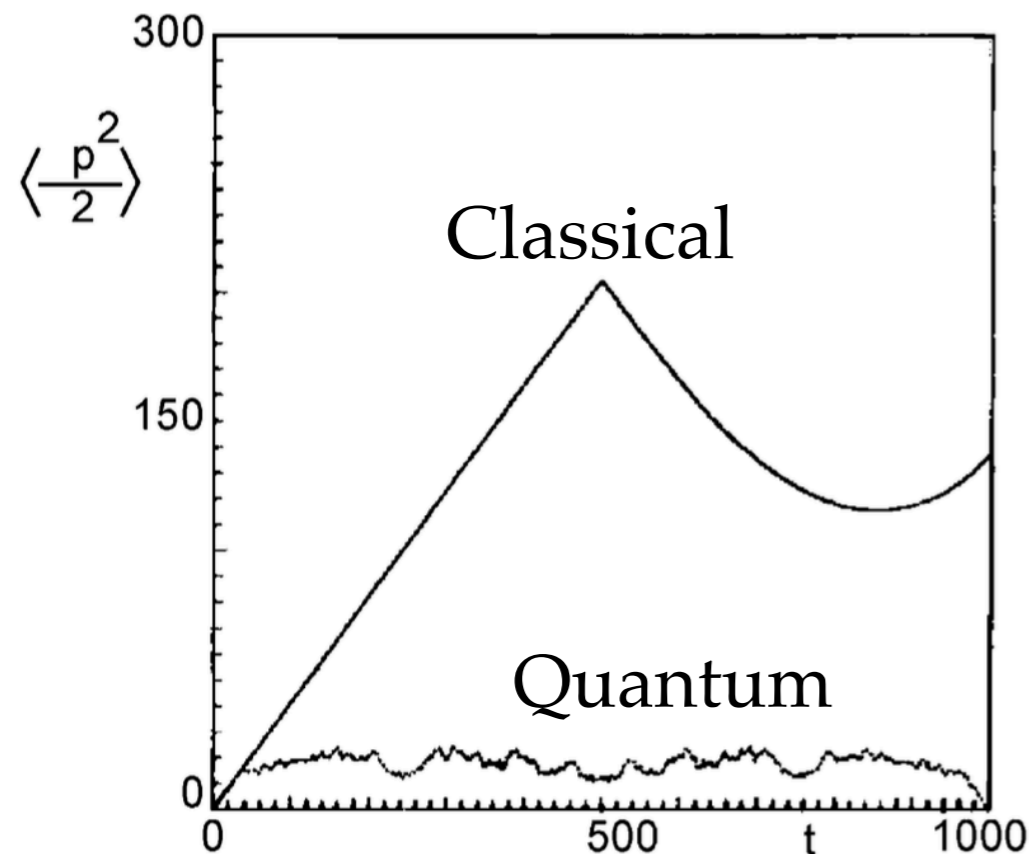


NOT SO FAST: ABSENCE OF QUANTUM CHAOS IN THE ROTOR

Lets begin with single particle models with canonical quantization

Quantum kicked rotor
$$\hat{H}_{KR} = \frac{\hat{p}^2}{2I} + k \cos(\hat{\theta}) \sum_{n=-\infty}^{+\infty} \delta(t - nT)$$

$$U_{KR} = e^{-i\hat{p}^2 T/2I} e^{-ik \cos(\hat{\theta})}$$



Classical model has diffusion

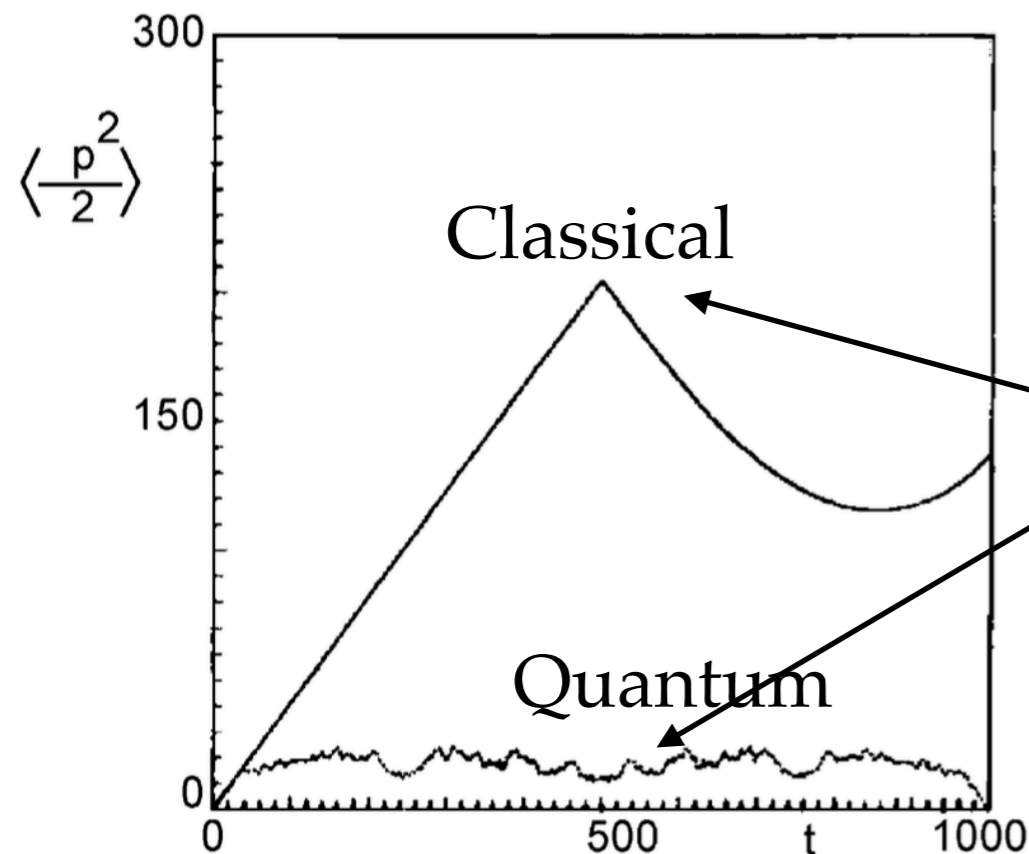
Quantum model undergoes dynamical localization

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Classical model has diffusion

Motion is reversed, exponentially sensitive
Only in the classical limit

Quantum model undergoes
dynamical localization

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$$U_{KR} = e^{-i\hat{p}^2 T/2I} e^{-ik \cos(\hat{\theta})}$$

We wish to transform this eigenvalue equation $U|\phi_j\rangle = e^{-iE_j T} |\phi_j\rangle$

NOT SO FAST: ABSENCE OF QUANTUM CHAOS IN THE ROTOR

Floquet operator $U_{KR} = e^{-i\hat{p}^2 T/2I} e^{-ik \cos(\hat{\theta})}$

Eigenvalue equation $U|\phi_j\rangle = e^{-iE_j T} |\phi_j\rangle$

$$e^{-i(\hat{p}^2/2 - E_j)T} e^{-ik \cos(\hat{\theta})} |\phi_j\rangle = |\phi_j\rangle$$

Define a new vector $|\tilde{\phi}_j\rangle = e^{ik \cos(\hat{\theta})} |\phi_j\rangle = e^{-iT(\hat{p}^2/2 - E_j)} |\phi_j\rangle$

Decompose the Floquet operator $e^{ix} = \frac{1 + i \tan(x/2)}{1 - i \tan(x/2)}$

$$e^{-ik \cos(\hat{\theta})} = \frac{1 + iW(\hat{\theta})}{1 - iW(\hat{\theta})}$$

NOT SO FAST: ABSENCE OF QUANTUM CHAOS IN THE ROTOR

$$|\tilde{\phi}_j\rangle = e^{ik \cos(\hat{\theta})} |\phi_j\rangle = e^{iT(\hat{p}^2/2 - E_j)} |\phi_j\rangle$$

$$|\phi_j\rangle = e^{-ik \cos(\hat{\theta})} |\tilde{\phi}_j\rangle = \frac{1 + iW(\hat{\theta})}{1 - iW(\hat{\theta})} |\tilde{\phi}_j\rangle$$

$$\frac{1}{1 + iW(\hat{\theta})} |\phi_j\rangle = \frac{1}{1 - iW(\hat{\theta})} |\tilde{\phi}_j\rangle \equiv |u_j\rangle$$

$$|\phi_j\rangle = (1 + iW(\hat{\theta})) |u_j\rangle \quad |\tilde{\phi}_j\rangle = (1 - iW(\hat{\theta})) |u_j\rangle = e^{iT(\hat{p}^2/2 - E_j)} |\phi_j\rangle$$

$$(1 - iW(\hat{\theta})) |u_j\rangle = e^{iT(\hat{p}^2/2 - E_j)} (1 + iW(\hat{\theta})) |u_j\rangle$$

NOT SO FAST: ABSENCE OF QUANTUM CHAOS IN THE ROTOR

$$(1 - iW(\hat{\theta}))|u_j\rangle = e^{iT(\hat{p}^2/2 - E_j)}(1 + iW(\hat{\theta}))|u_j\rangle$$

In the theta basis $u_j(\theta) = \langle\theta|u\rangle$

$$(1 - iW(\theta))u_j(\theta) = e^{iT((-i\partial_\theta)^2/2 - E_j)}(1 + iW(\theta))u_j(\theta)$$

Fourier transforming to the momentum basis $\hat{p}|m\rangle = m|m\rangle$

$$W(\theta) = \sum_{m=-\infty}^{\infty} W_m e^{im\theta} \quad u_j(\theta) = \sum_{m=-\infty}^{\infty} u_j^m e^{im\theta} \quad u_j^m = \langle m|u\rangle$$

$$T_m u_j^m + \sum_{r \neq 0} W_r u_j^{m+r} = -W_0 u_j^m$$

NOT SO FAST: ABSENCE OF QUANTUM CHAOS IN THE ROTOR

We have mapped the quantum dynamics to a discrete Schrodinger equation in momentum space

$$T_m u_j^m + \sum_{r \neq 0} W_r u_j^{m+r} = -W_0 u_j^m$$

On-site energies

$$T_m = \tan[(E_j - m^2/2)/2]$$

Long-range
Hopping between site
 m and r

$$W_r = - \int_0^{2\pi} \frac{d\theta}{2\pi} e^{ir\theta} \tan(k \cos \theta/2)$$

Energy eigenvalue

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Pseudorandom numbers

The sequence $(E_j - m^2/2) \bmod \pi$

Is ergodic in the interval $[0, \pi]$

Density of levels $\rho(T_m) dT_m = dE_j / \pi$

$$W_r = - \int_0^{2\pi} \frac{d\theta}{2\pi} e^{ir\theta} \tan(k \cos \theta / 2)$$

$$\frac{dT_m}{dE_j} = \frac{1}{2} \cos^{-2}((E_j - m^2/2)/2) = \frac{1}{2} (1 + T_m^2)$$

$$\rho(T_m) = \frac{1}{\pi(1 + T_m^2)} \quad \text{(Normalizing)}$$

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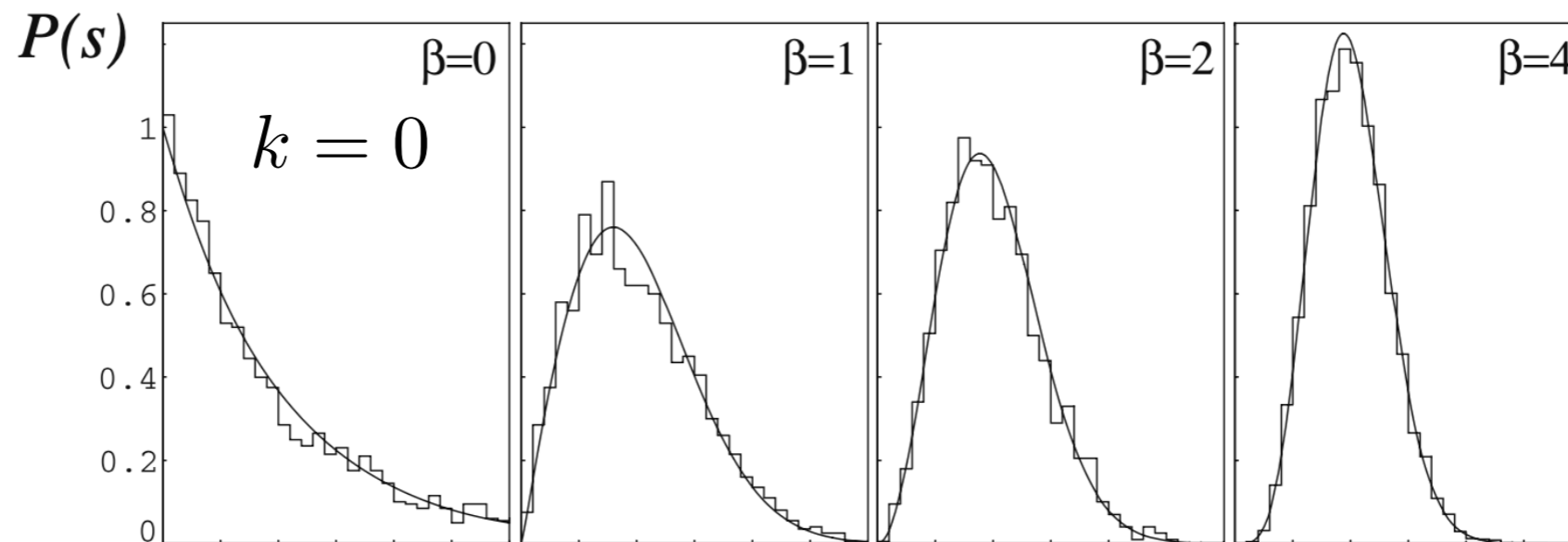
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This is a disordered 1D hopping model, Anderson localizes!

SUMMARY SO FAR

What we have learned so far:

The kicked top serves as a quintessential model for quantum chaos, the 3 symmetry classes, GOE, GUE, GSE can be probed



Quantum Signatures of Chaos, Haake

The kicked rotor becomes dynamically localized and does not represent chaotic behavior.

What about the quantum spin chain?

QUANTUM SPIN CHAINS

Focusing on the antiferromagnetic Heisenberg spin chain

$$H = J \sum_i \hat{\mathbf{S}}_i \cdot \hat{\mathbf{S}}_{i+1}$$

It turns out to be **integrable**, solvable using the Bethe ansatz

Simply taking the quantum limit does not yield a chaotic system.

Add next nearest neighbor coupling

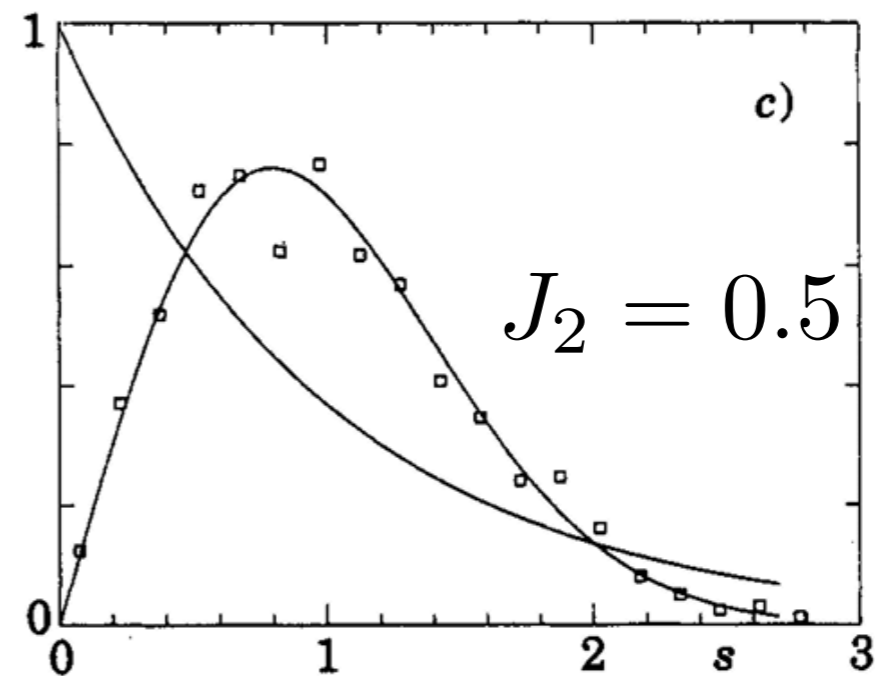
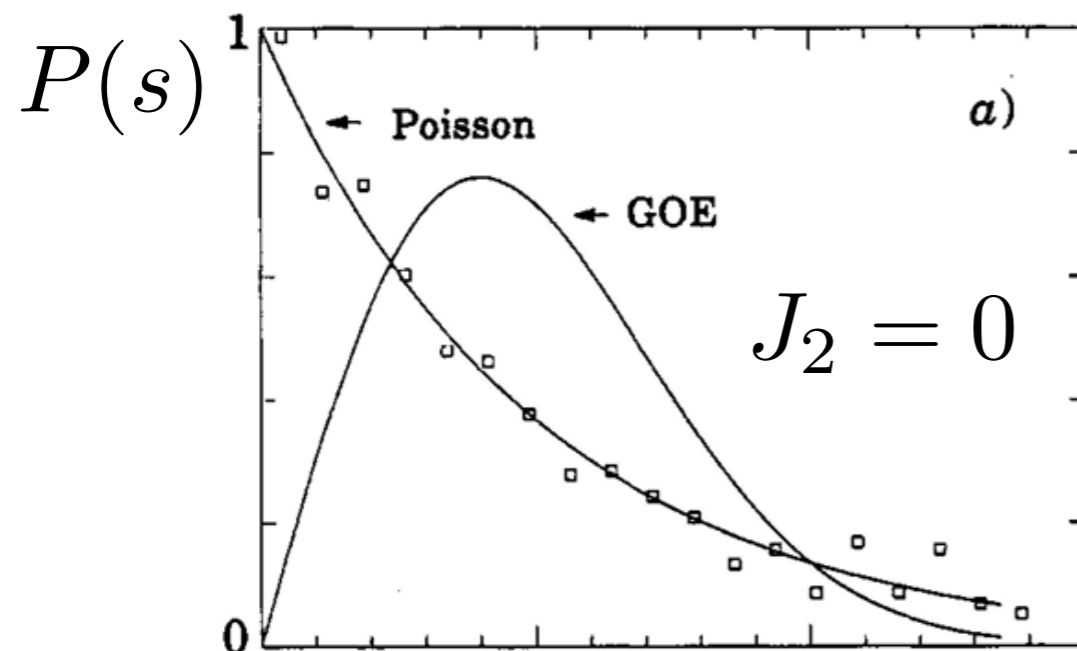
$$H = J_1 \sum_i \hat{\mathbf{S}}_i \cdot \hat{\mathbf{S}}_{i+1} + J_2 \sum_i \hat{\mathbf{S}}_i \cdot \hat{\mathbf{S}}_{i+2}$$

Do the many-body levels now look chaotic?

QUANTUM SPIN CHAINS

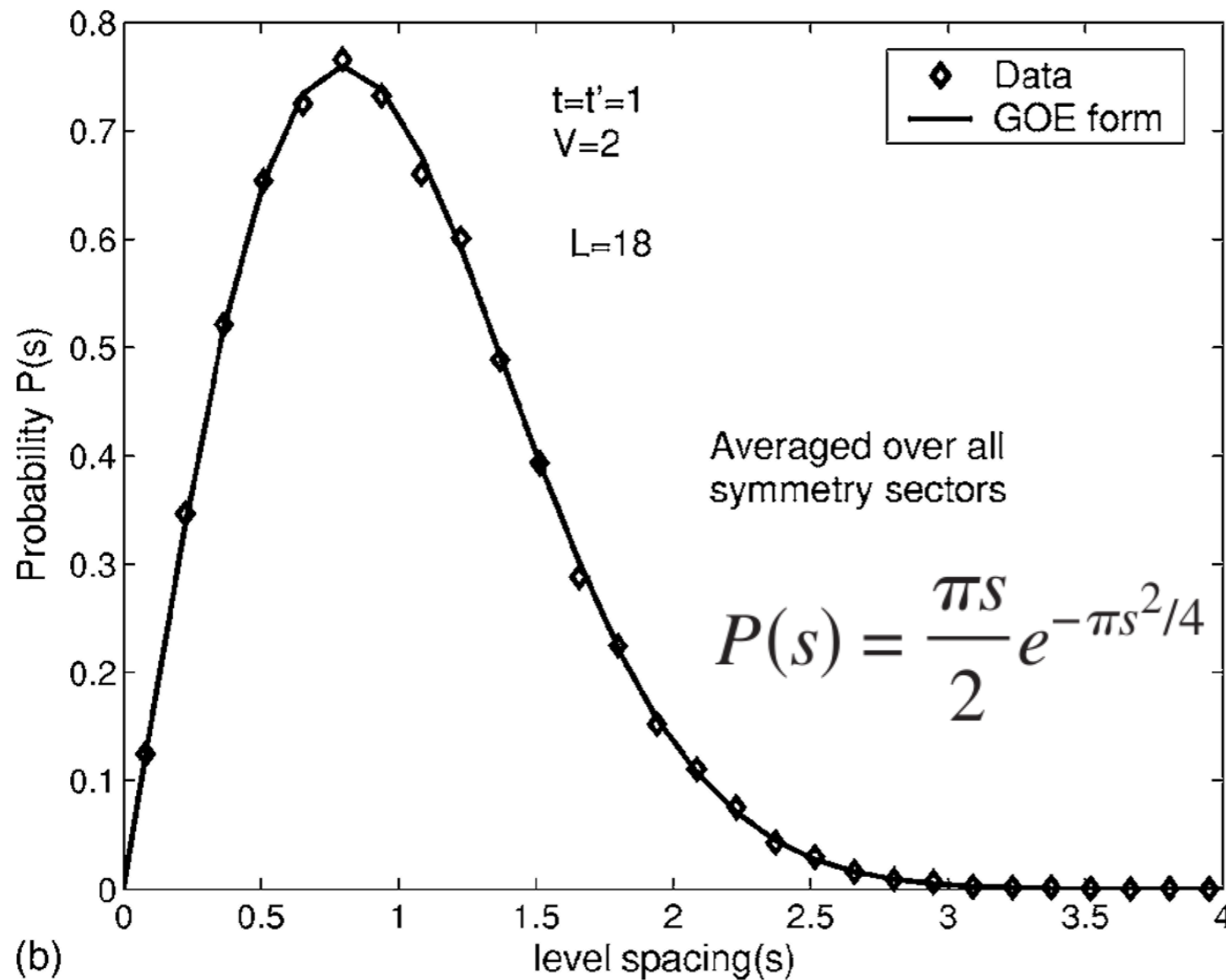
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Away from integrability the many-body levels satisfy random matrix theory



STRONGLY INTERACTING 1D FERMIONS

$$H = -t \sum_j c_j^+ c_{j+1} - t' \sum_j c_j^+ c_{j+2} + \text{h.c.} + V \sum_j n_j n_{j+1}$$



The many-body
energy levels

$$s_i = E_i - E_{i-1}$$

Well described using
the GOE ensemble

NEED METHODS TO DESCRIBE MANY-BODY QUANTUM CHAOS

Random matrix theory is useful but its not enough.

Many-body quantum chaotic systems should thermalize:

How is this described?

How does this take place?

In addition to correlations, we will use the entanglement entropy (and several other notions from quantum information)

CHARACTERIZING QUANTUM MANY BODY SYSTEMS

A useful characterization for quantum many-body systems beyond correlations is the nature of their **entanglement structure**

$$\langle \Psi | \mathcal{O}(x, t) \mathcal{O}(0, 0) | \Psi \rangle$$

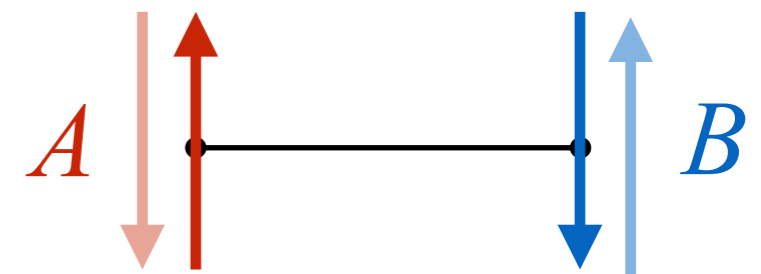
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Example: 2 spin-1/2s in a singlet

$$|\psi\rangle = \frac{1}{\sqrt{2}} (|\uparrow_A\rangle |\downarrow_B\rangle - |\downarrow_A\rangle |\uparrow_B\rangle)$$



“1 bit of entanglement”

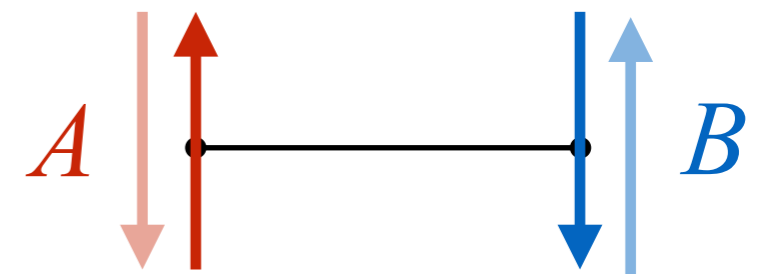
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Von Neumann entanglement entropy

$$S_E = -\text{Tr}_A(\rho_A \log \rho_A)$$

$$\rho = |\psi\rangle\langle\psi|$$

$$\rho_A = \text{Tr}_B(|\psi\rangle\langle\psi|)$$

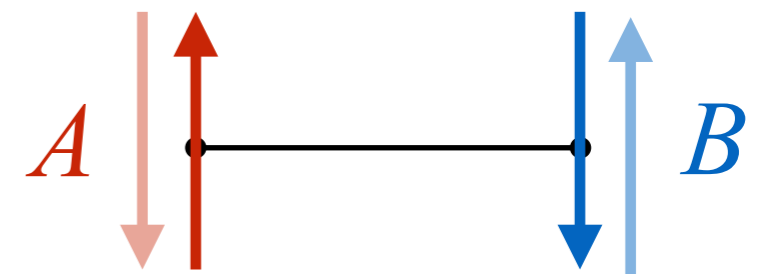
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Von Neumann entanglement entropy

$$S_E = -\text{Tr}_A(\rho_A \log \rho_A) \\ = \log 2$$

“1 bit of entanglement”

$$\rho = |\psi\rangle \langle \psi|$$

$$\rho_A = \text{Tr}_B(|\psi\rangle \langle \psi|)$$

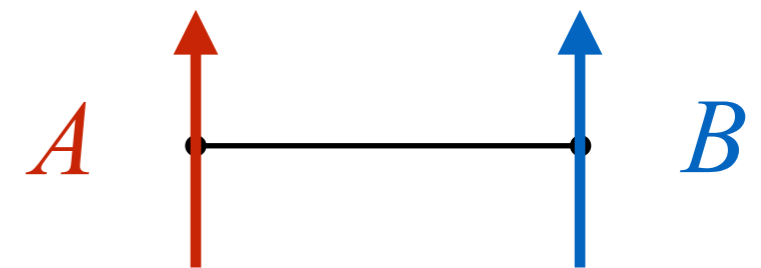
$$= \begin{pmatrix} \frac{1}{2} & 0 \\ 0 & \frac{1}{2} \end{pmatrix}$$

CHARACTERIZING QUANTUM MANY BODY SYSTEMS

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Example: 2 spin-1/2s aligned



$$|\psi\rangle = |\uparrow_A\rangle |\uparrow_B\rangle$$

Von Neumann entanglement entropy

$$\begin{aligned} S_E &= -\text{Tr}_A(\rho_A \log \rho_A) \\ &= 0 \end{aligned}$$

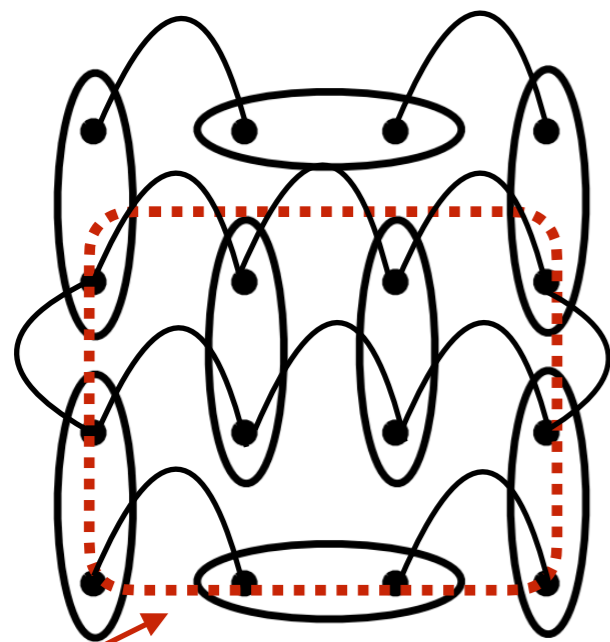
No entanglement!

$$\rho = |\psi\rangle \langle \psi|$$

$$\begin{aligned} \rho_A &= \text{Tr}_B(|\psi\rangle \langle \psi|) \\ &= \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \end{aligned}$$

CHARACTERIZING QUANTUM MANY BODY SYSTEMS

Now consider a many body problem
e.g. a large number of spin-1/2s interacting



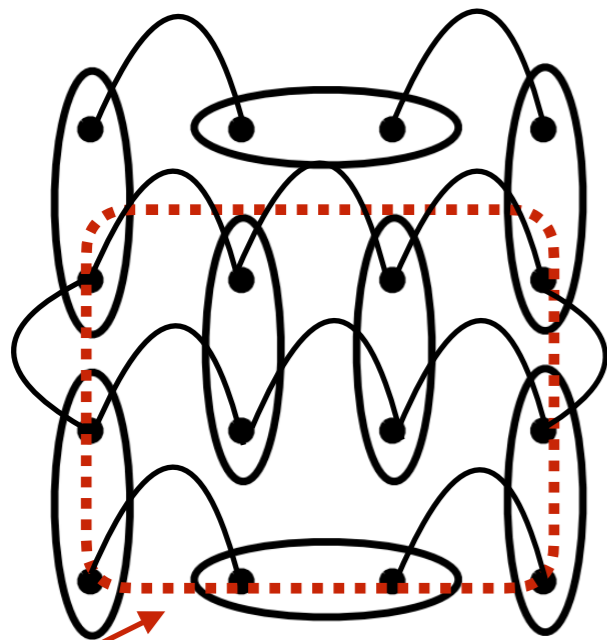
perimeter of length L

a weakly entangled state
short range collection of singlets:
e.g. ground state of a valence bond solid

$$S_E \sim L^{d-1} \text{ "area law"}$$

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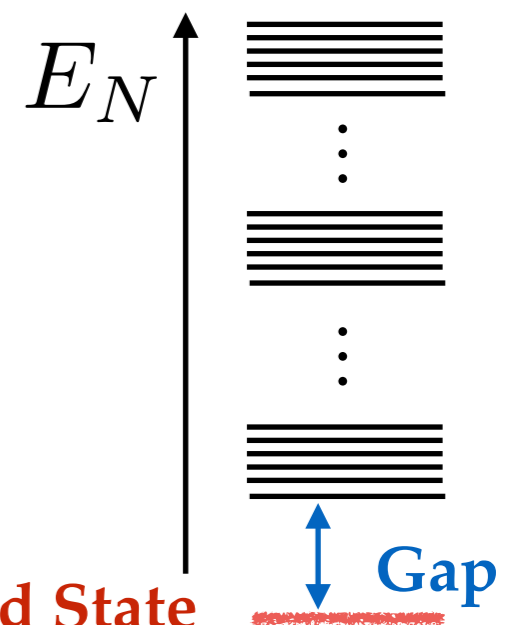
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characteristic of the ground state
entanglement of gapped Hamiltonians

proven in 1D, Hasings JSTAT (2007)

many-body
Energy spectra

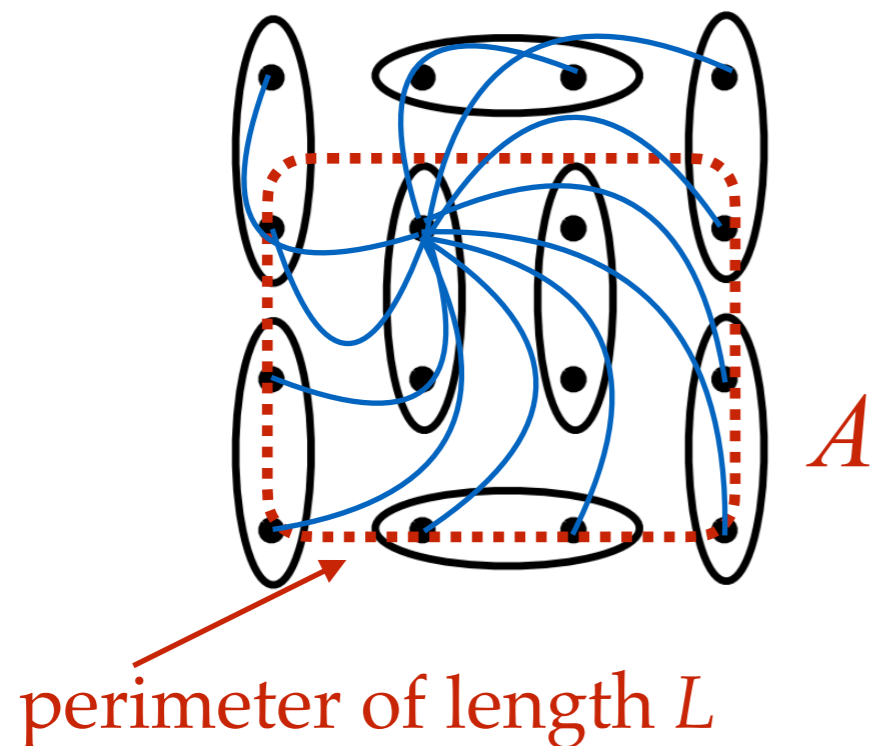


Ground State

Gap

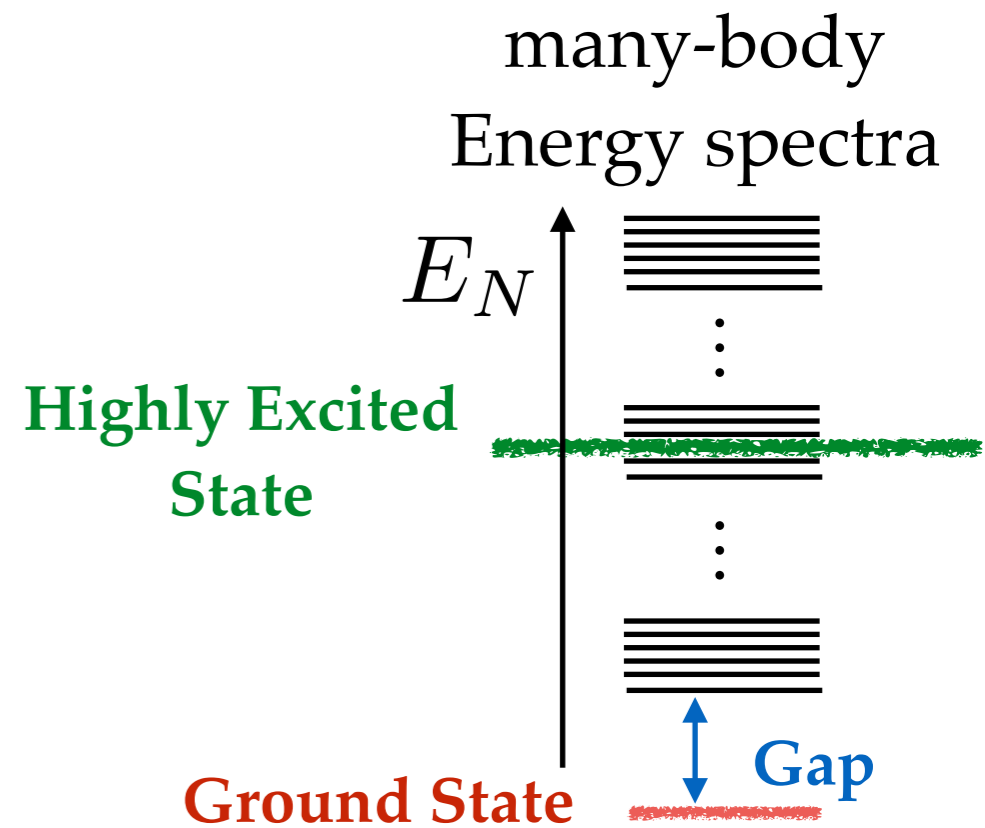
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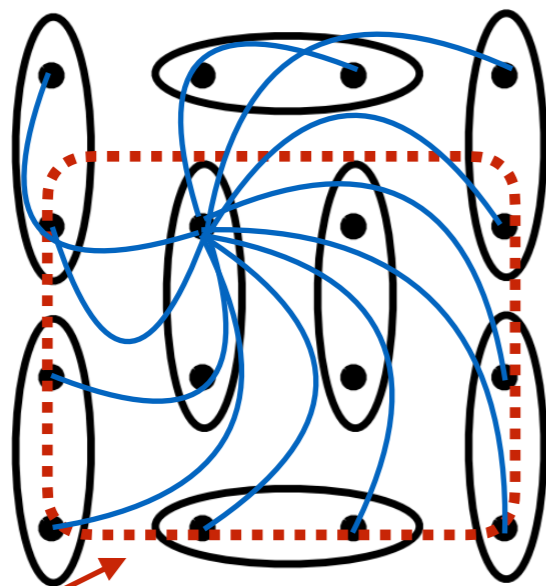
A highly entangled state
at **finite energy density**

$$S_E \sim L^d \quad \text{"volume law"}$$



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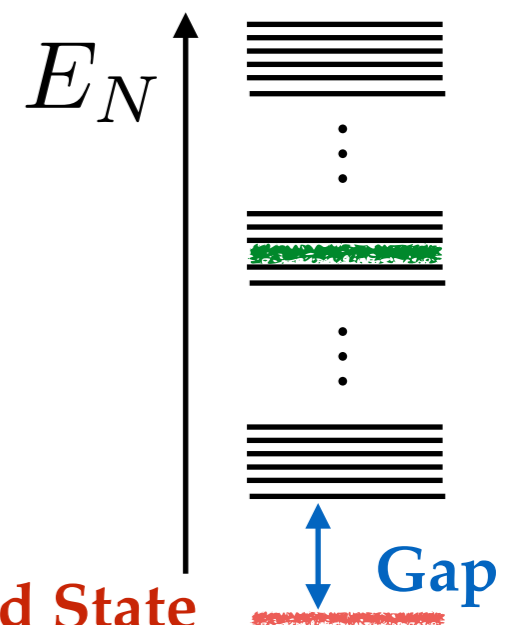
perimeter of length L

The long time-dynamics of
a many body Hamiltonian

$$|\Psi_0\rangle = |\uparrow\downarrow \dots \uparrow\downarrow\rangle$$

$$|\Psi(t)\rangle = e^{-iHt} |\Psi_0\rangle$$

many-body
Energy spectra



CHARACTERIZING QUANTUM MANY BODY SYSTEMS

In summary, quantum states can be broadly classified by their entanglement structure.

weakly entangled $S_E \sim L^{d-1}$ “area law”

highly entangled $S_E \sim L^d$ “volume law”

Many-body quantum chaotic states are highly entangled

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weakly entangled $S_E \sim L^{d-1}$ “area law”

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Many-body quantum chaotic states are highly entangled

In $d=1$, critical states acquire a log-correction

$$S_E \sim \log L$$

Calabrese and Cardy, J. Stat. Mech (2004)

OUTLINE

- I. Lecture series layout
- II. Classical chaos
- III. Quantum chaos
- IV. Quantum thermalization**
- V. Evading thermalization

THERMALIZATION IN CLASSICAL SYSTEMS

Classical thermalization takes place through the “ergodic hypothesis”

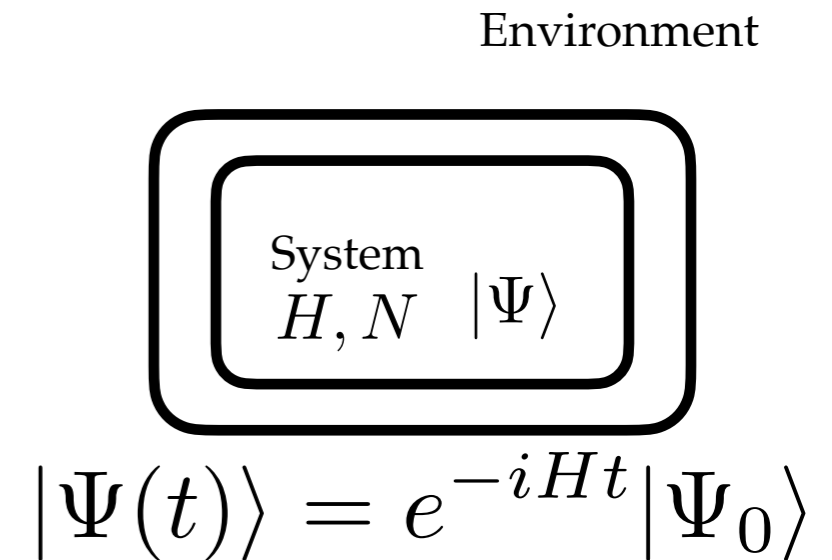
Ergodic systems can explore all of their phase space.

Chaotic systems need not be ergodic, and ergodic systems
need not be chaotic.

FROM MANY-BODY QUANTUM CHAOS TO THERMALIZATION

How does an isolated quantum system reach thermal equilibrium?

When a system has thermalized it can be described by a few parameters, e.g. chemical potential and temperature



Deutsch PRA (1991) Srednicki PRE (1994) Rigol, Dunjko, and Olshanii, Nature (2008) Linden, et al PRE (2009)

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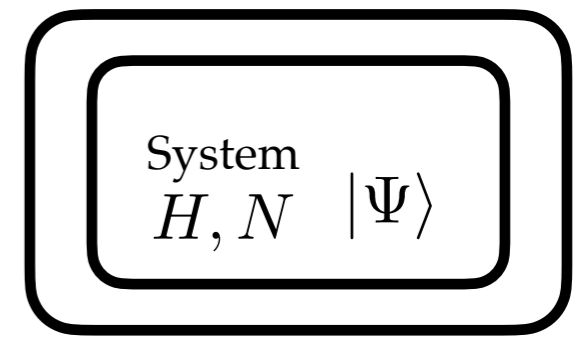
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How does an isolated quantum system reach thermal equilibrium?

When a system has thermalized it can be described by a few parameters, e.g. chemical potential and temperature

But Hamiltonian time evolution is unitary, and it cannot erase information, (only picks up phases)

Environment



$$|\Psi(t)\rangle = e^{-iHt} |\Psi_0\rangle$$

All of the information of the initial state must be preserved in the evolution. It appears to be a paradox....

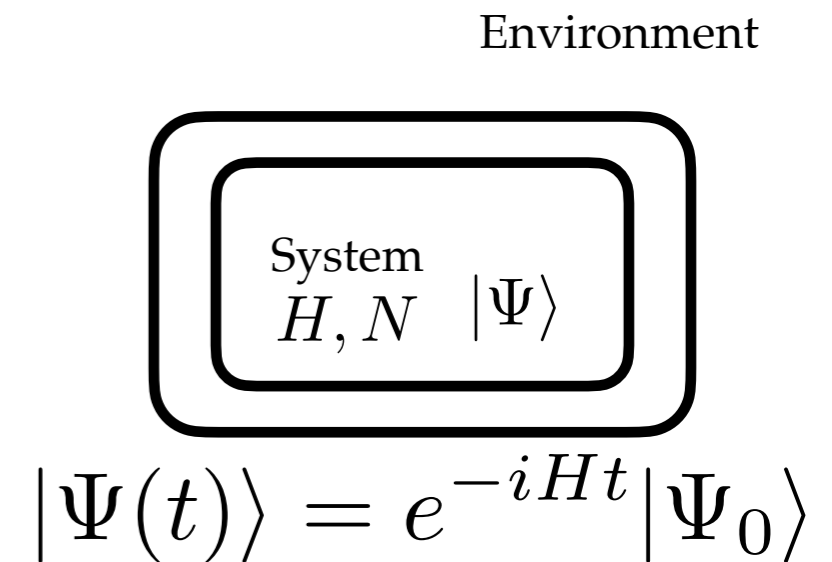
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FROM MANY-BODY QUANTUM CHAOS TO THERMALIZATION

How does an isolated quantum system reach thermal equilibrium?

The resolution to this putative paradox
Is that the unitary evolution has spread the
local information about the initial state
across the entire system.



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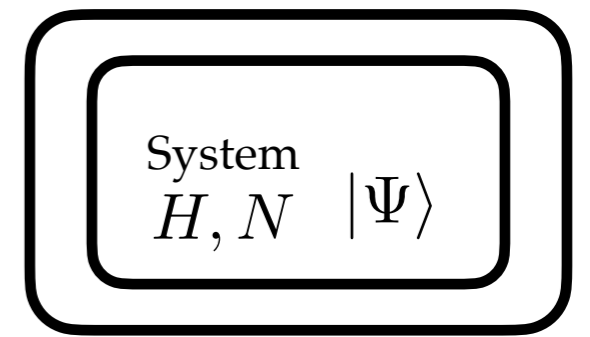
How does an isolated quantum system reach thermal equilibrium?

The resolution to this putative paradox
Is that the unitary evolution has spread the
local information about the initial state
across the entire system.

For this to be possible it has to be
sufficiently entangled and locally chaotic.

At late times, this local information has
been hidden and cannot be retrieved by
local measurements.

Environment



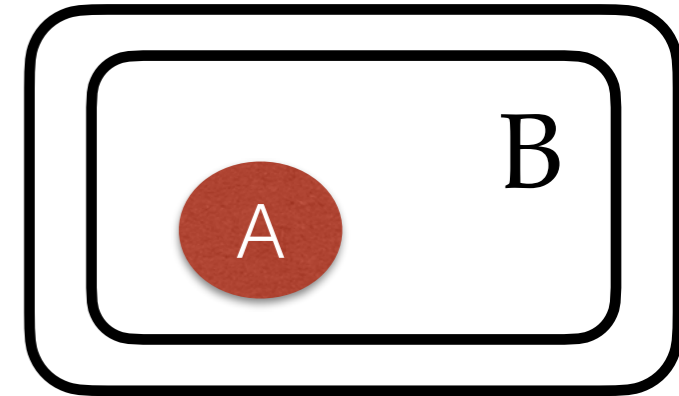
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EIGENSTATE THERMALIZATION HYPOTHESIS (ETH)

Lets now focus on local observables in a small region A and ask if region A can thermalize while region B acts like a reservoir



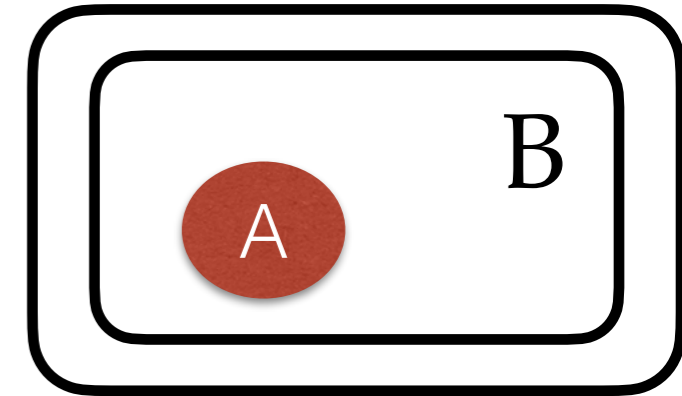
Consider an initial state with weight over a large number of energy eigenstates

$$|\Psi(0)\rangle = \sum_n c_n |n\rangle$$

$$|\Psi(t)\rangle = \sum_n c_n e^{-iE_n t} |n\rangle$$

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$$|\Psi(0)\rangle = \sum_n c_n |n\rangle$$

For a local operator (in A)

$$|\Psi(t)\rangle = \sum_n c_n e^{-iE_n t} |n\rangle$$

$$\lim_{t \rightarrow \infty} \frac{1}{t} \langle \Psi(t) | \hat{O} | \Psi(t) \rangle = \sum_n A_{nn} O_{nn} \equiv \langle \hat{O} \rangle_{\text{eq}}$$

such a steady state has “equilibrated”

Linden, Popescu, Short, Winter PRE (2009)

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Linden, Popescu, Short, Winter PRE (2009)

For the system to have thermalized, ETH requires

$$\langle \hat{O} \rangle_{\text{MC}} = \frac{1}{\mathcal{N}} \sum_{|E_n - E| \leq \Delta} O_{nn} \approx \langle \Psi(E) | \hat{O} | \Psi(E) \rangle \quad \langle \hat{O} \rangle_{\text{eq}} = \langle \hat{O} \rangle_{\text{MC}}$$

Micro-canonical ensemble

Rigol, Dunjko, and Olshanii, Nature (2008)

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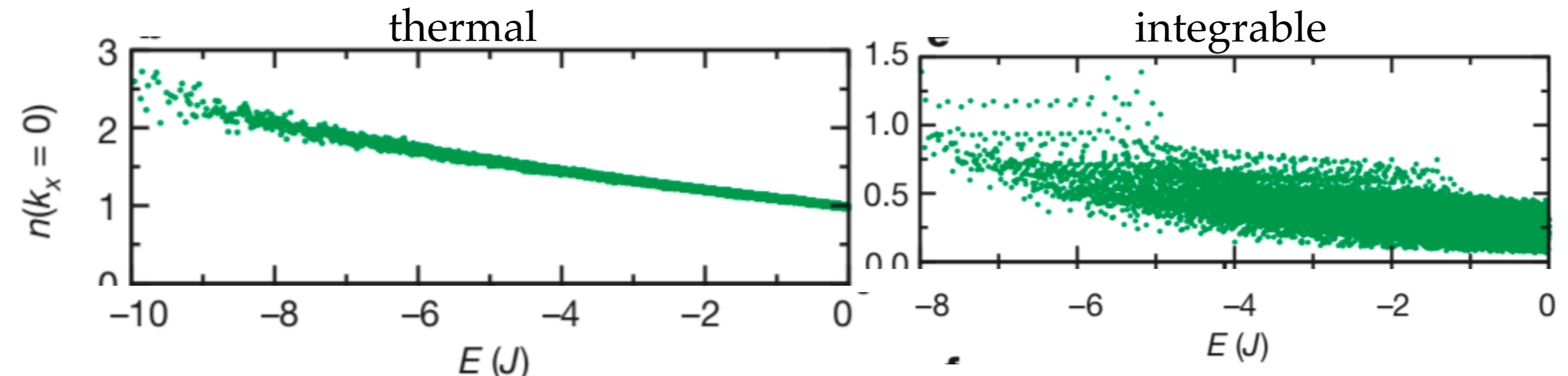
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$$\begin{aligned} O(t) &\equiv \langle \psi(t) | \hat{O} | \psi(t) \rangle = \sum_{m,n} C_m^* C_n e^{i(E_m - E_n)t} O_{mn} \\ &= \sum_m |C_m|^2 O_{mm} + \sum_{m,n \neq m} C_m^* C_n e^{i(E_m - E_n)t} O_{mn} \end{aligned}$$

ETH can also be stated as an asatz for matrix elements

$$\langle \hat{O} \rangle_{\text{eq}} = \langle \hat{O} \rangle_{\text{MC}} \quad O_{mn} = O(\bar{E}) \delta_{mn} + e^{-S(\bar{E})/2} f_O(\bar{E}, \omega) R_{mn}$$

A STRONGER VERSION OF ETH

For a local operator

$$\lim_{t \rightarrow \infty} \frac{1}{t} \langle \Psi(t) | \hat{O} | \Psi(t) \rangle = \sum_n A_{nn} O_{nn} \equiv \langle \hat{O} \rangle_{\text{eq}} \quad \text{such a steady state has "equilibrated"}$$

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$$\langle \hat{O} \rangle_{\text{C}} = \frac{1}{Z} \text{Tr}(e^{-\beta H} \hat{O}) \quad \text{Canonical ensemble}$$

ETH requires

$$\langle \hat{O} \rangle_{\text{eq}} = \langle \hat{O} \rangle_{\text{C}} = \langle \hat{O} \rangle_{\text{MC}}$$

temperature is set by the avg energy

$$\langle \Psi | H | \Psi \rangle = \text{Tr}(H e^{-\beta H}) / \text{Tr}(e^{-\beta H})$$

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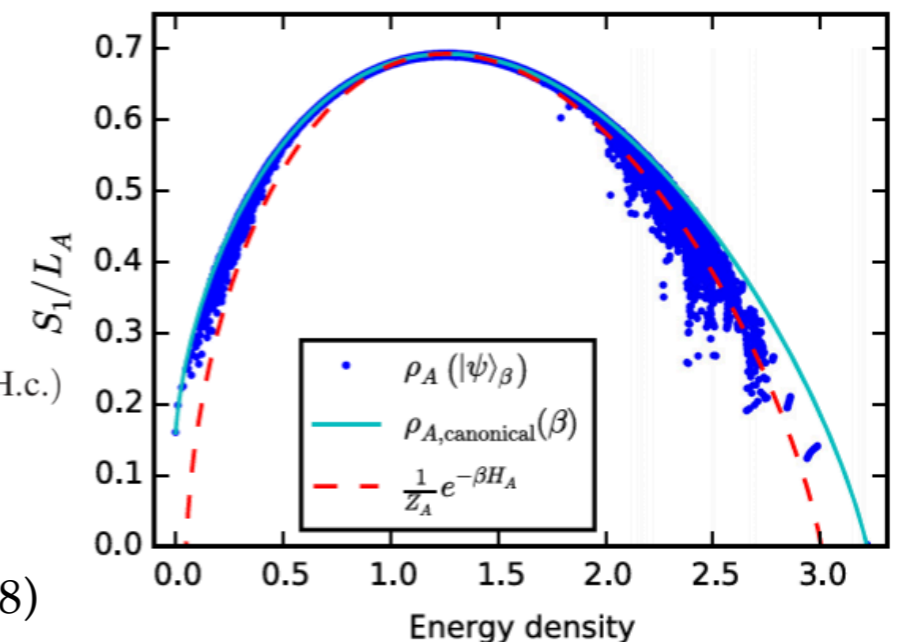
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$$\langle \Psi | H | \Psi \rangle = \text{Tr}(H e^{-\beta H}) / \text{Tr}(e^{-\beta H})$$

Then we can conclude

$$\rho_A \sim \text{Tr}_B(e^{-\beta H})$$

$$H = -\sum_i (t b_i^\dagger b_{i+1} + t' b_i^\dagger b_{i+2} + \text{H.c.}) + \sum_i (V n_i n_{i+1} + V' n_i n_{i+2}),$$

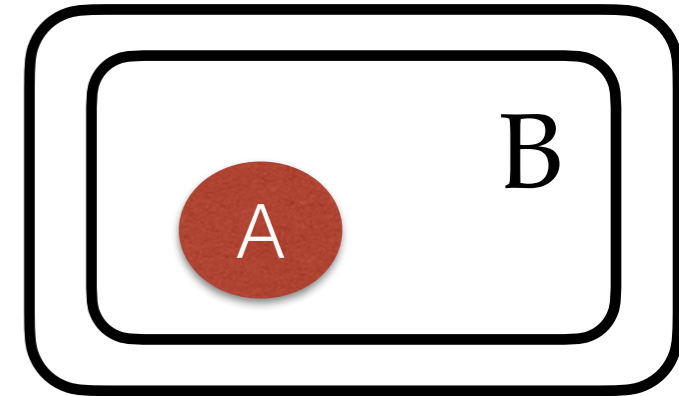


AN EVEN STRONGER VERSION OF ETH

Conventional thermalization

$$\rho_A \sim \text{Tr}_B(e^{-\beta H})$$

Canonical
ensemble



Deep thermalization: Consider making an extensive number of local measurements in B. The wave function in A is projected and weighted by the Born probability of the measurement, and the reduced density matrix in A is now:

Projected ensemble: The ensemble of all possible states in A after the measurement and Born probabilities in B

$$\rho_A^{DT} = \sum_{\mathbf{m}} p_{\mathbf{m}} |\psi_A^{(\mathbf{m})}\rangle \langle \psi_A^{(\mathbf{m})}|$$

Born probability of measurement outcome in B Wavefunction in A after the measurement in B

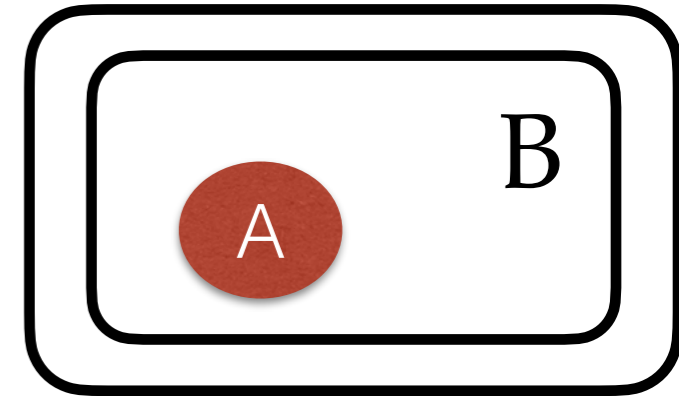
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Born probability of
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Wavefunction in A after
the measurement in B



The projected ensemble has the full distribution of the reduced density matrix

AN EVEN STRONGER VERSION OF ETH

Projected ensemble: The ensemble of all possible states in A after the measurement and Born probabilities

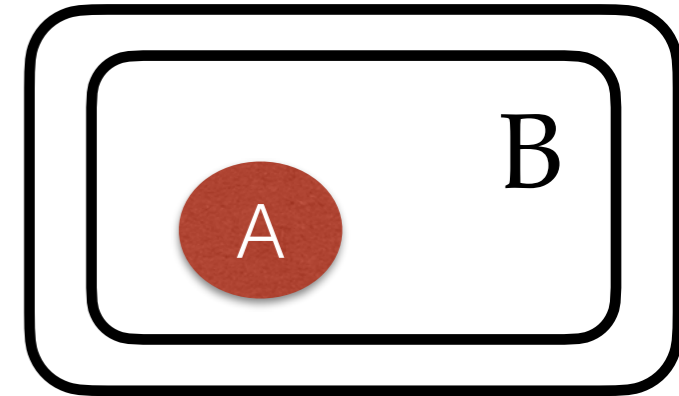
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Born probability of measurement outcome in B

Wavefunction in A after the measurement in B

Higher moments can be estimated from it.

$$(\rho_A^{DT})^k = \sum_{\mathbf{m}} p_{\mathbf{m}} |\psi_A^{(\mathbf{m})}\rangle \langle \psi_A^{(\mathbf{m})}|^{\otimes k}$$



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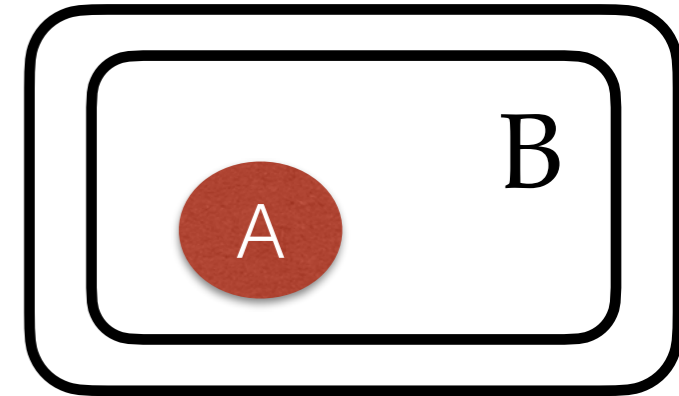
In the absence of any symmetries or conservation laws, at infinite temperature

The mean reduces to ETH

$$(\rho_A^{DT})^{k=1} = \rho_A$$

$$\rho^{(k)}(t) \rightarrow \rho_H^{(k)}$$

$$\rho_H^{(k)} = \int_{\psi \sim \text{Haar}(\mathcal{H}_A)} d\psi (|\psi\rangle \langle \psi|)^{\otimes k}$$



The projected ensemble has the full distribution of the reduced density matrix

WHAT ABOUT THE PROCESS OF THERMILIZATION

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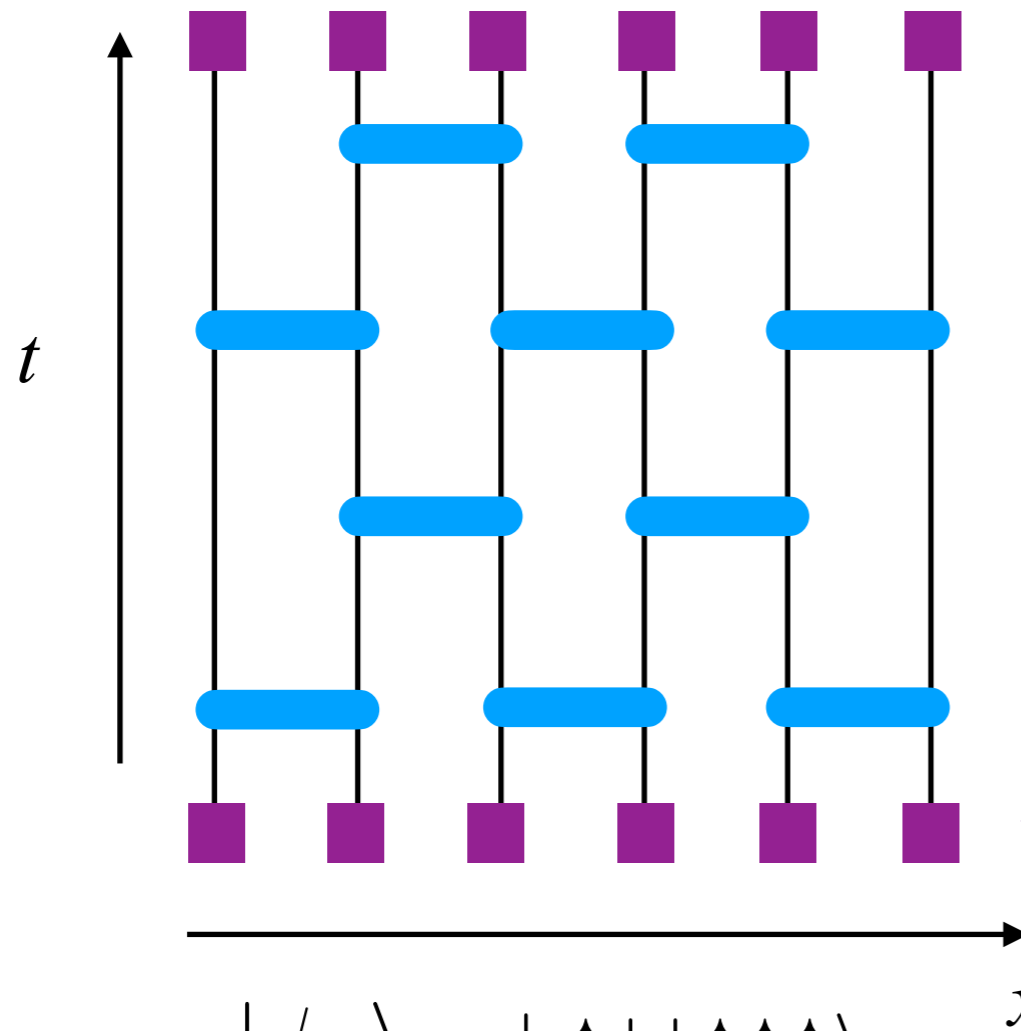
Must capture entanglement growth and information
propagation.

Random unitary circuits have emerged
as minimal models to address this question.

von Keyserlingk, Rakovszky, Pollmann, and Sondhi PRX (2018)

A. Nahum, S. Vijay, and J. Haah PRX (2018)

RANDOM UNITARY CIRCUITS



q -states “spins”
e.g. $q=2$ is a spin-1/2 chain

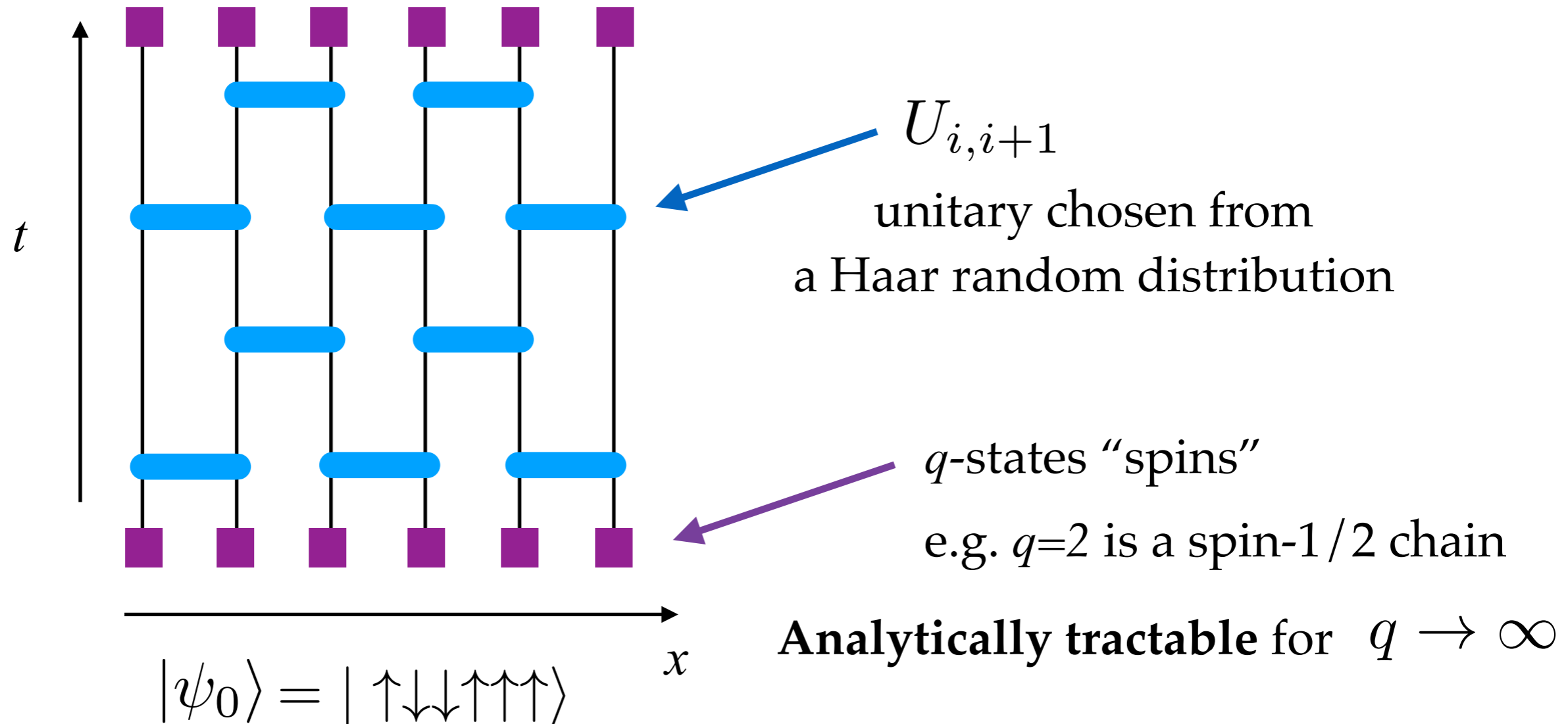
Analytically tractable for $q \rightarrow \infty$

e.g. some random product state

A. Nahum, S. Vijay, and J. Haah PRX (2018)

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RANDOM UNITARY CIRCUITS

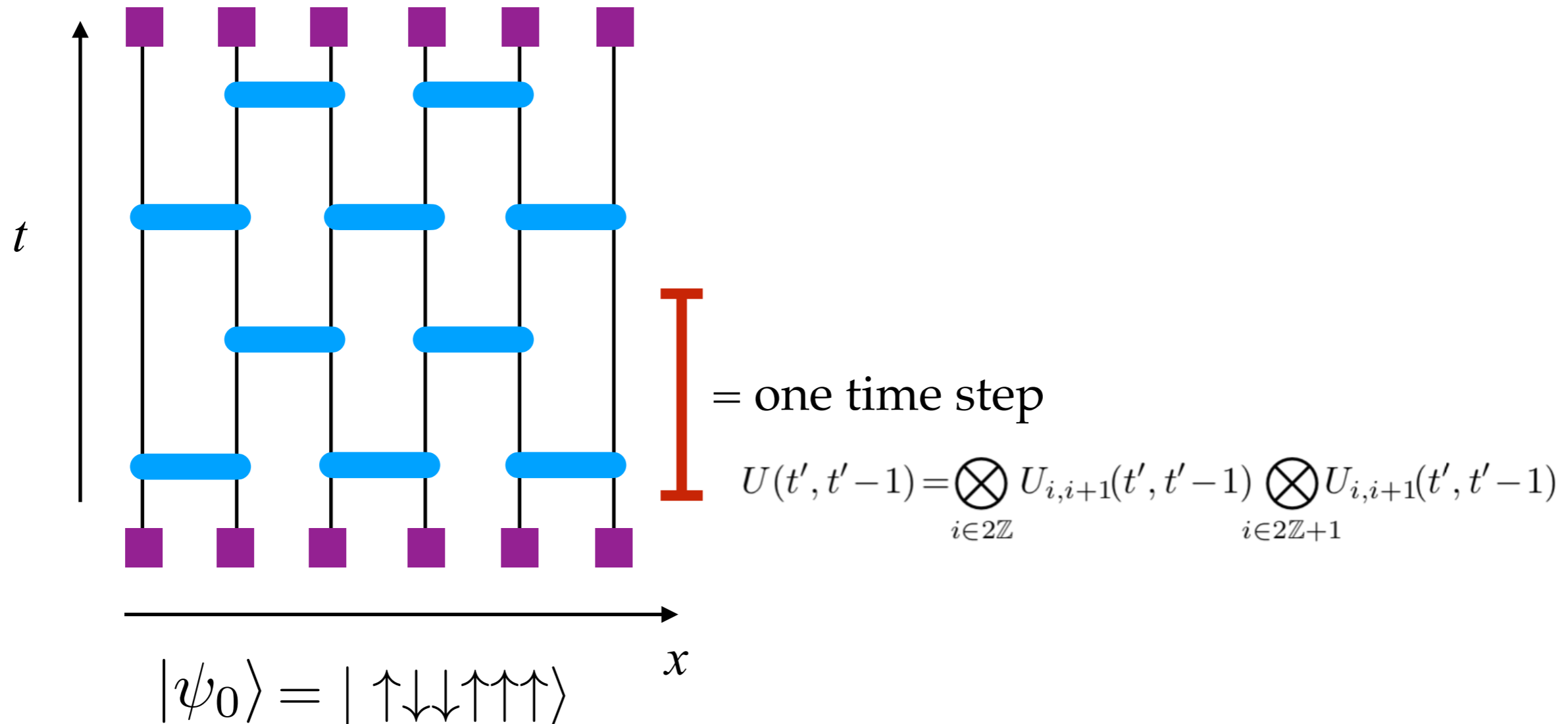


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A. Nahum, S. Vijay, and J. Haah PRX (2018)

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RANDOM UNITARY CIRCUITS

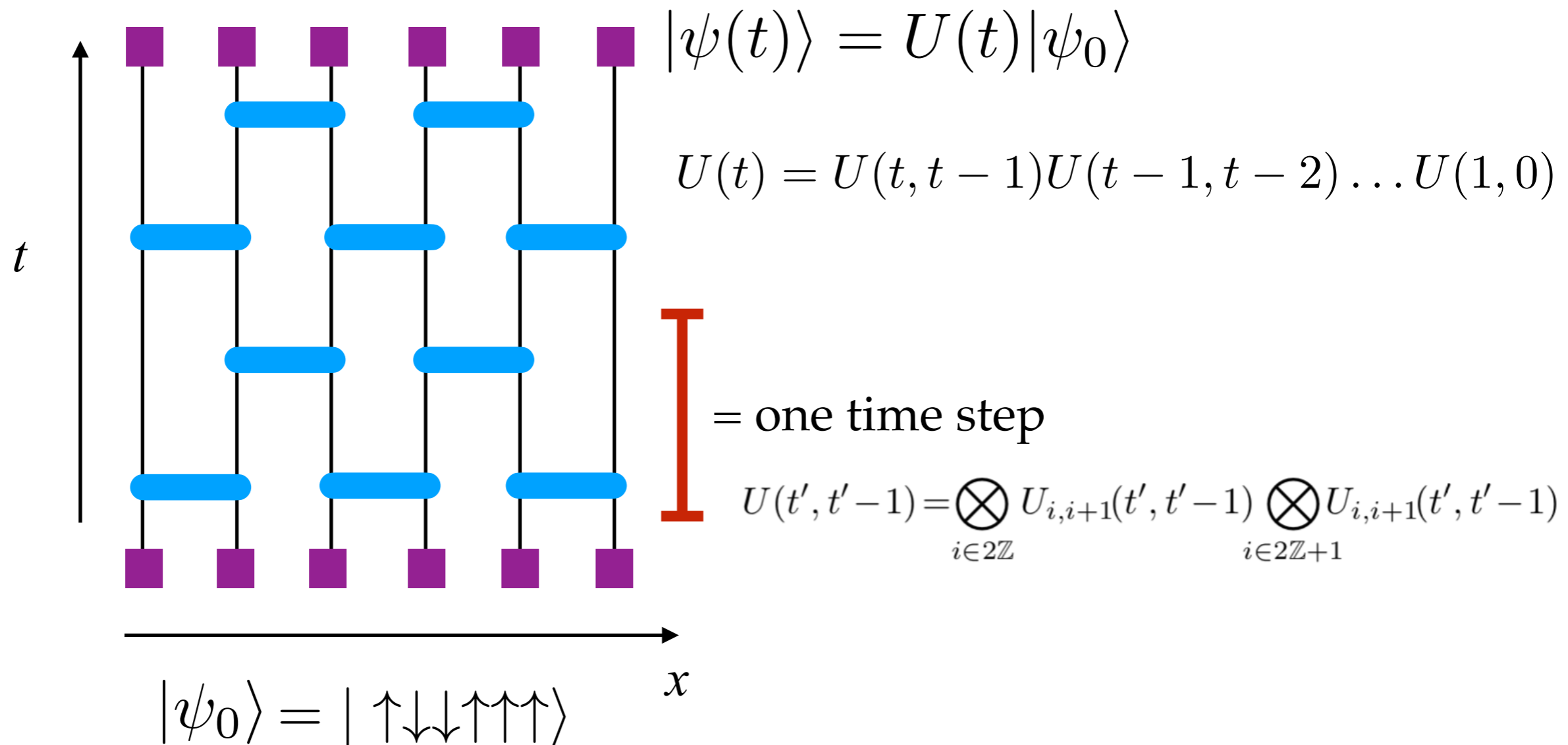


e.g. some random product state

A. Nahum, S. Vijay, and J. Haah PRX (2018)

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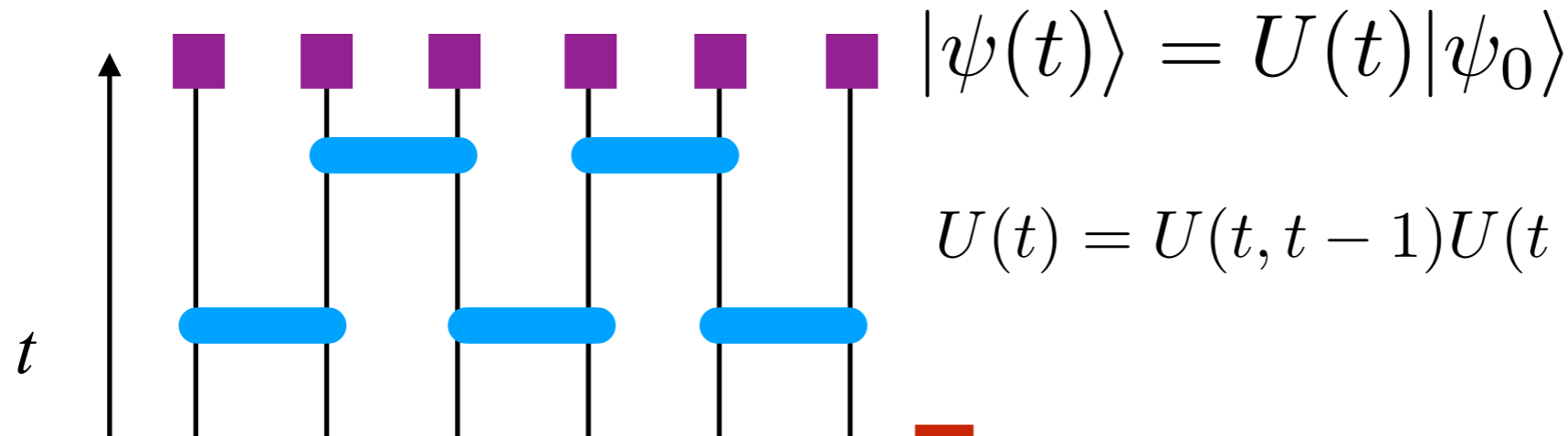


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von Keyserlingk, Rakovszky, Pollmann, and Sondhi PRX (2018)

RANDOM UNITARY CIRCUITS



It is expected / argued that the behavior found here is generic to many-body quantum dynamics

the time step

$$U(t', t'-1) = \bigotimes_{i \in 2\mathbb{Z}} U_{i, i+1}(t', t'-1) \bigotimes_{i \in 2\mathbb{Z}+1} U_{i, i+1}(t', t'-1)$$

$$|\psi_0\rangle = | \uparrow \downarrow \uparrow \downarrow \uparrow \uparrow \uparrow \rangle$$

e.g. some random product state

A. Nahum, S. Vijay, and J. Haah PRX (2018)

von Keyserlingk, Rakovszky, Pollmann, and Sondhi PRX (2018)

STATISTICS OF THE REDUCED DENSITY MATRIX

The reduced density matrix for a random pure state (i.e. Page state) $|\Psi\rangle = \sum_i \alpha_i |C_i\rangle$ random complex numbers

$\rho_A = \text{Tr}_B(|\Psi\rangle\langle\Psi|)$ is a **Wishart** random matrix

Znidaric, J. Phys. A: Math. Theor. (2007)

Yang et al PRL (2016)

Chen and Ludwig PRB (2018)

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Chen and Ludwig PRB (2018)

$P(\{Y_{iJ}\}) = \mathcal{N}^{-1} \exp\{-\frac{\beta}{2} N_B \text{Tr}(YY^\dagger)\}$ Y is a complex random matrix

$\hat{\rho}_A \equiv \frac{YY^\dagger}{\text{Tr}(YY^\dagger)}$ $W = YY^\dagger$ W is from the Wishart random matrix ensemble

STATISTICS OF THE REDUCED DENSITY MATRIX

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Captures states that satisfy **ETH** $\hat{\rho}_A \equiv \frac{YY^\dagger}{\text{Tr}(YY^\dagger)}$

$$W = YY^\dagger$$

STATISTICS OF THE REDUCED DENSITY MATRIX

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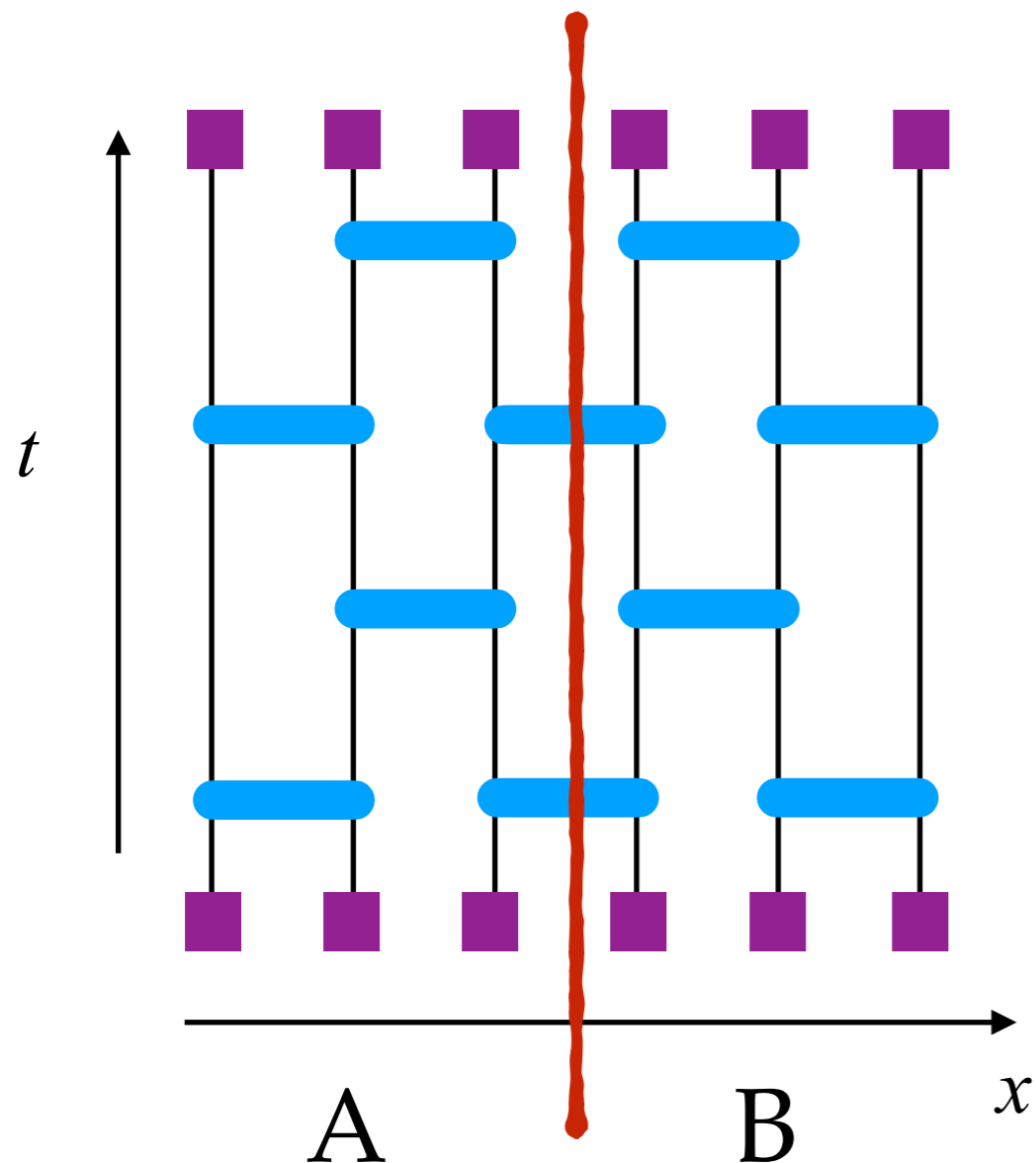
Chen and Ludwig PRB (2018)

And the density of levels follows the **Marchenko-Pastur** distribution $P(\lambda) \sim 1/\sqrt{\lambda}$

ENTANGLEMENT SPECTRA

Entanglement spectrum of the reduced density matrix is analogous to a set of “Energy levels”

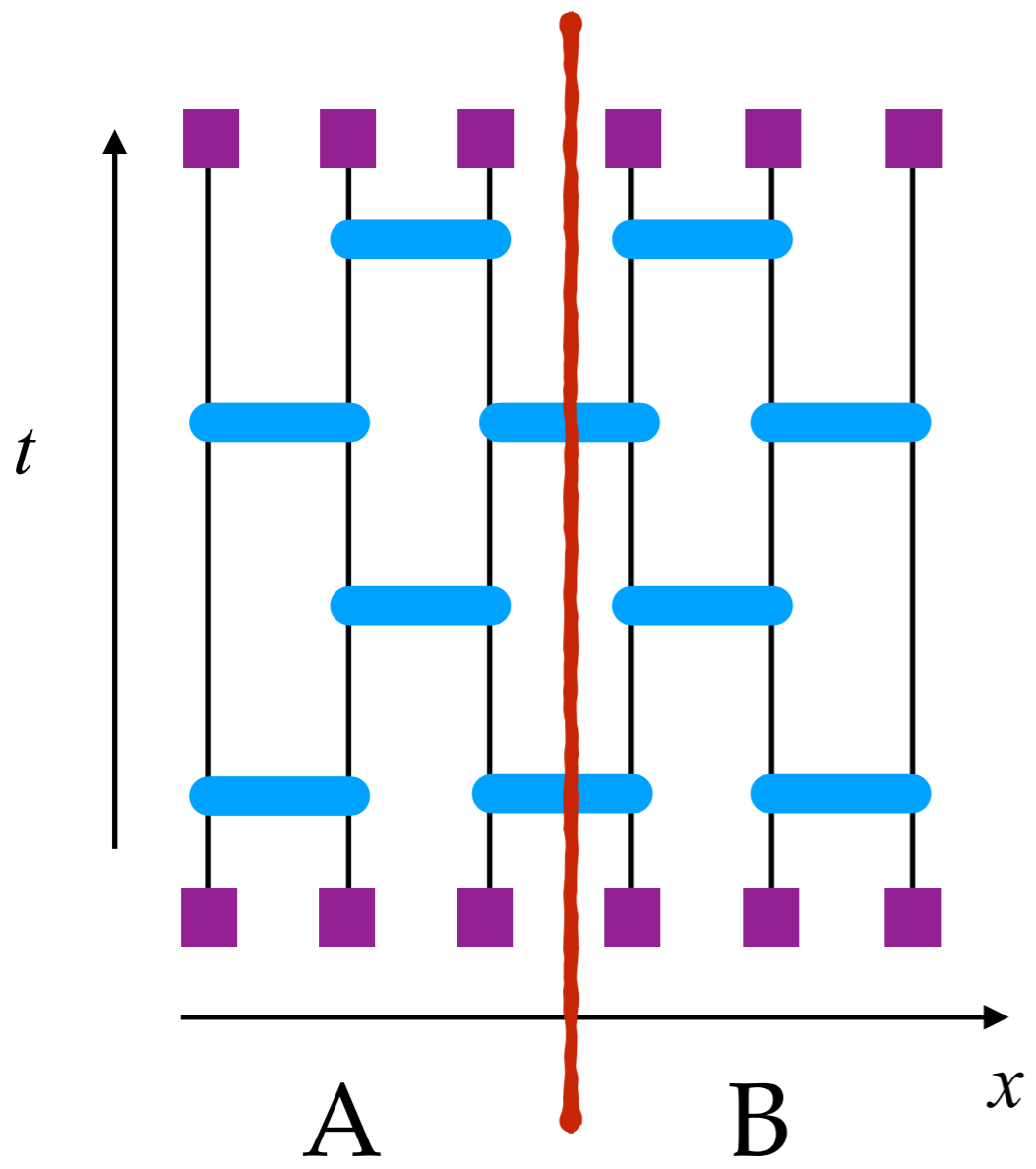
Li and Haldane PRL (2008)



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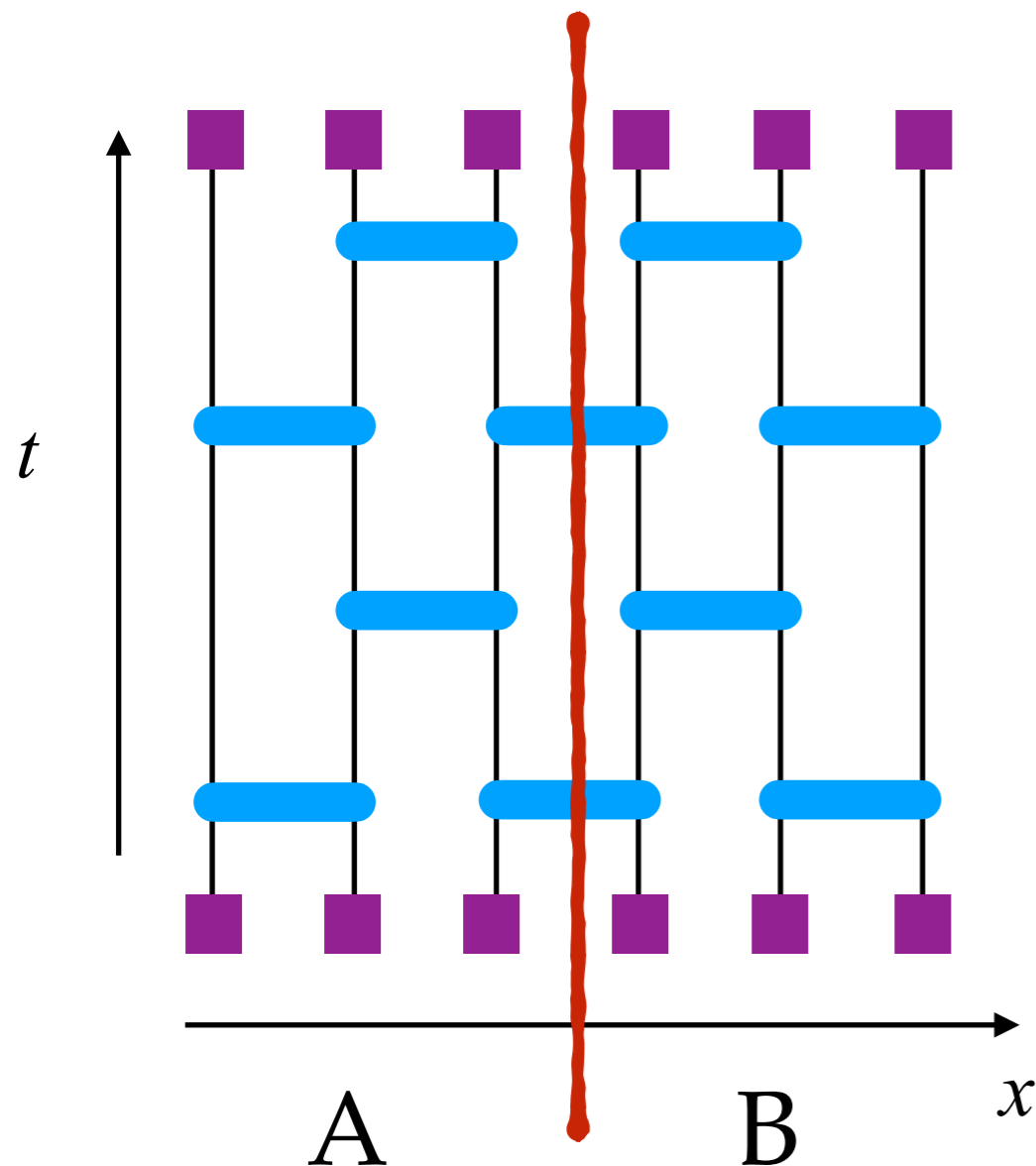
$$\hat{\rho}_A = \text{Tr}_B \left[\hat{U}(t) |\Psi_0\rangle \langle \Psi_0| \hat{U}^\dagger(t) \right]$$

Eigenvalues of the reduced density matrix $\{\lambda_n\}$

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Eigenvalues of the reduced density matrix $\{\lambda_n\}$

Entanglement spectra $\{E_n\} = \{-\log \lambda_n\}$

Renyi entropies $S_\alpha = \frac{1}{\alpha - 1} \log \left(\sum_n \lambda_n^\alpha \right)$

ENTANGLEMENT SPECTRA

Entanglement spectrum of the reduced density matrix is analogous to a set of “Energy levels”

Li and Haldane PRL (2008)

For Hamiltonian dynamics
in a system that satisfies ETH

$$\rho_A \sim \text{Tr}_B(e^{-\beta H})$$

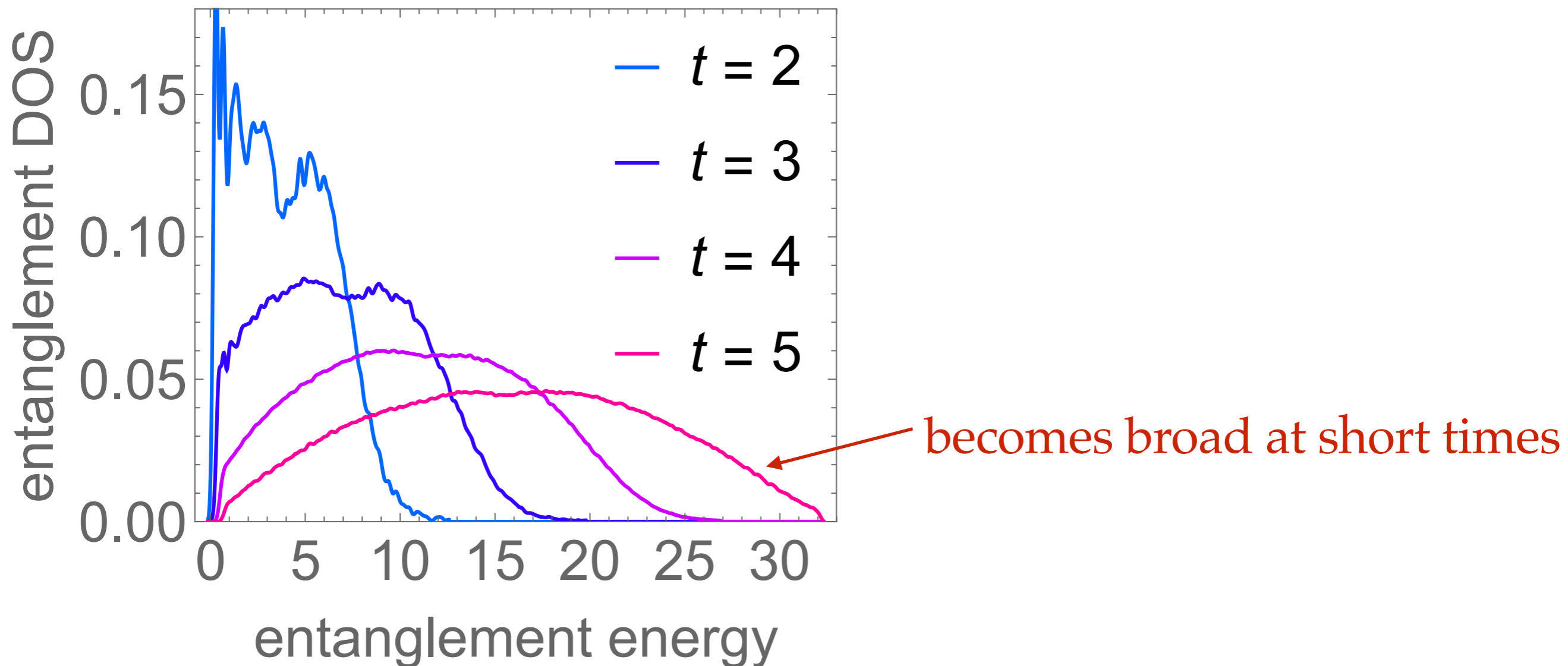
Thus the entanglement spectrum inherits the spectral statistics of the projected Hamiltonian (up to boundary terms)

SCRAMBLING AT SHORT TIMES

Starting from an initial product state,
the **rank** of the reduced density matrix
grows as a function of time

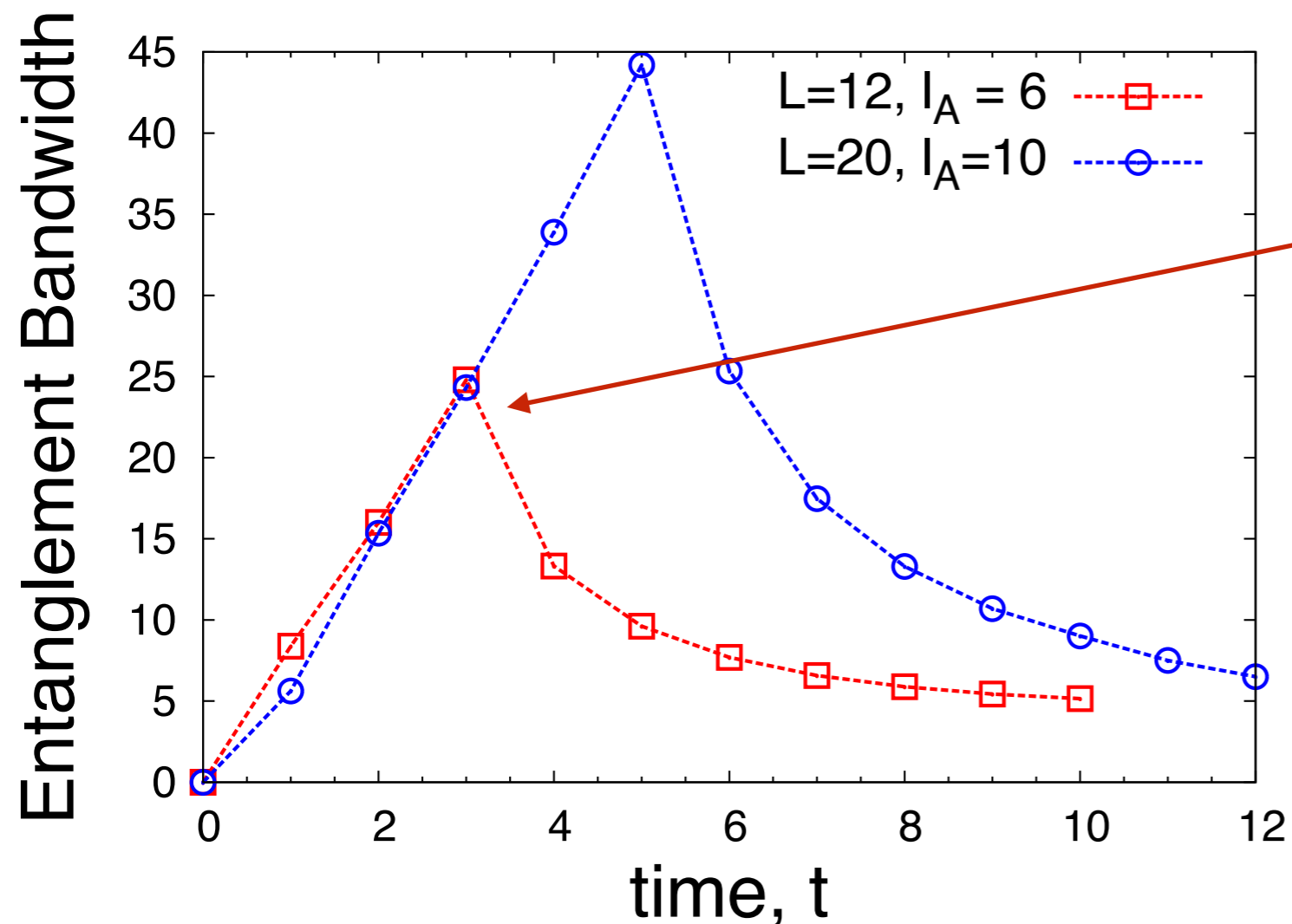
SCRAMBLING AT SHORT TIMES

Starting from an initial product state, the **rank** of the reduced density matrix grows as a function of time



ENTANGLEMENT ENTROPY VS BANDWIDTH

Growth of the rank of the reduced density matrix captured by the entanglement bandwidth!

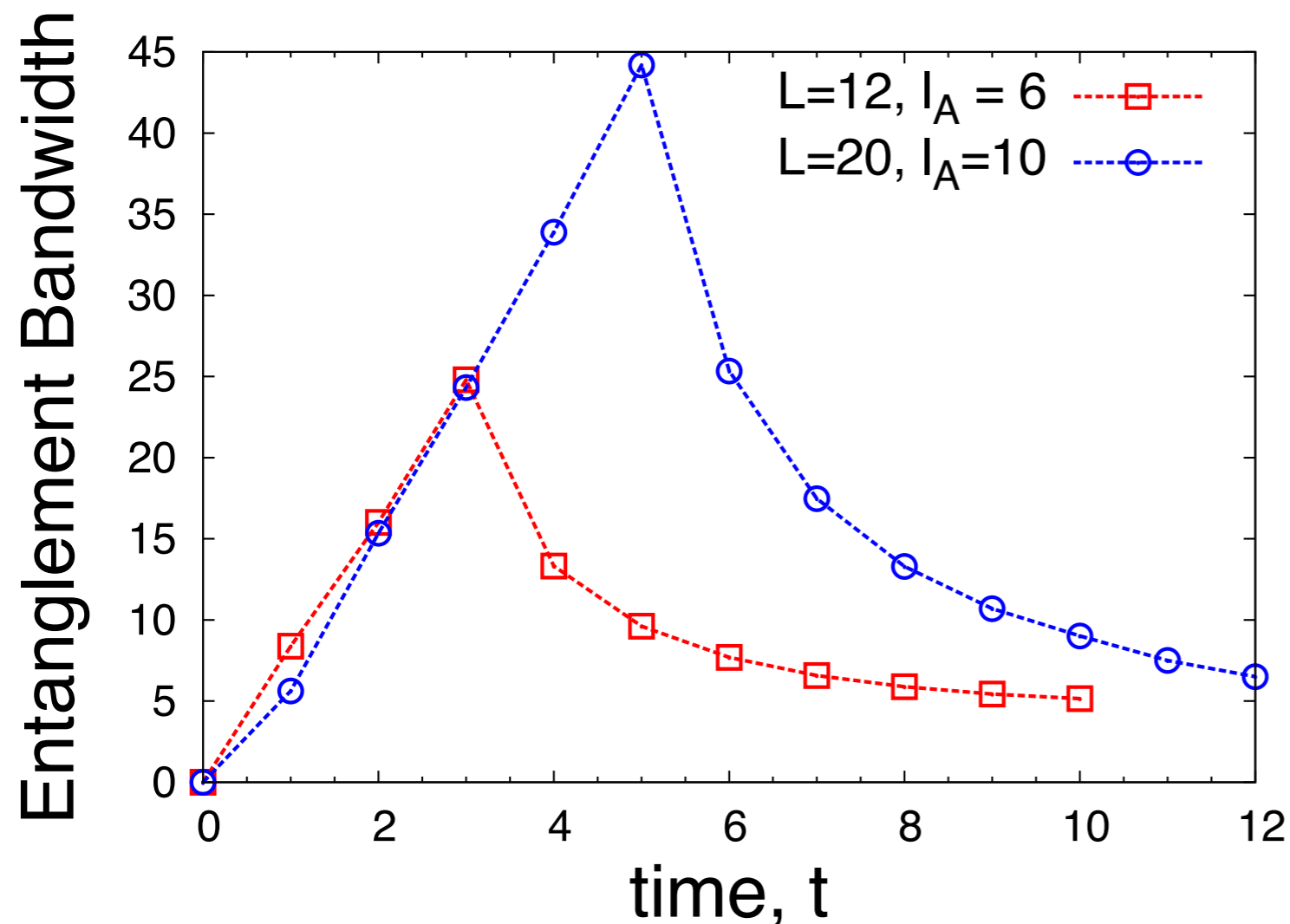


becomes broad at short times

ENTANGLEMENT ENTROPY VS BANDWIDTH

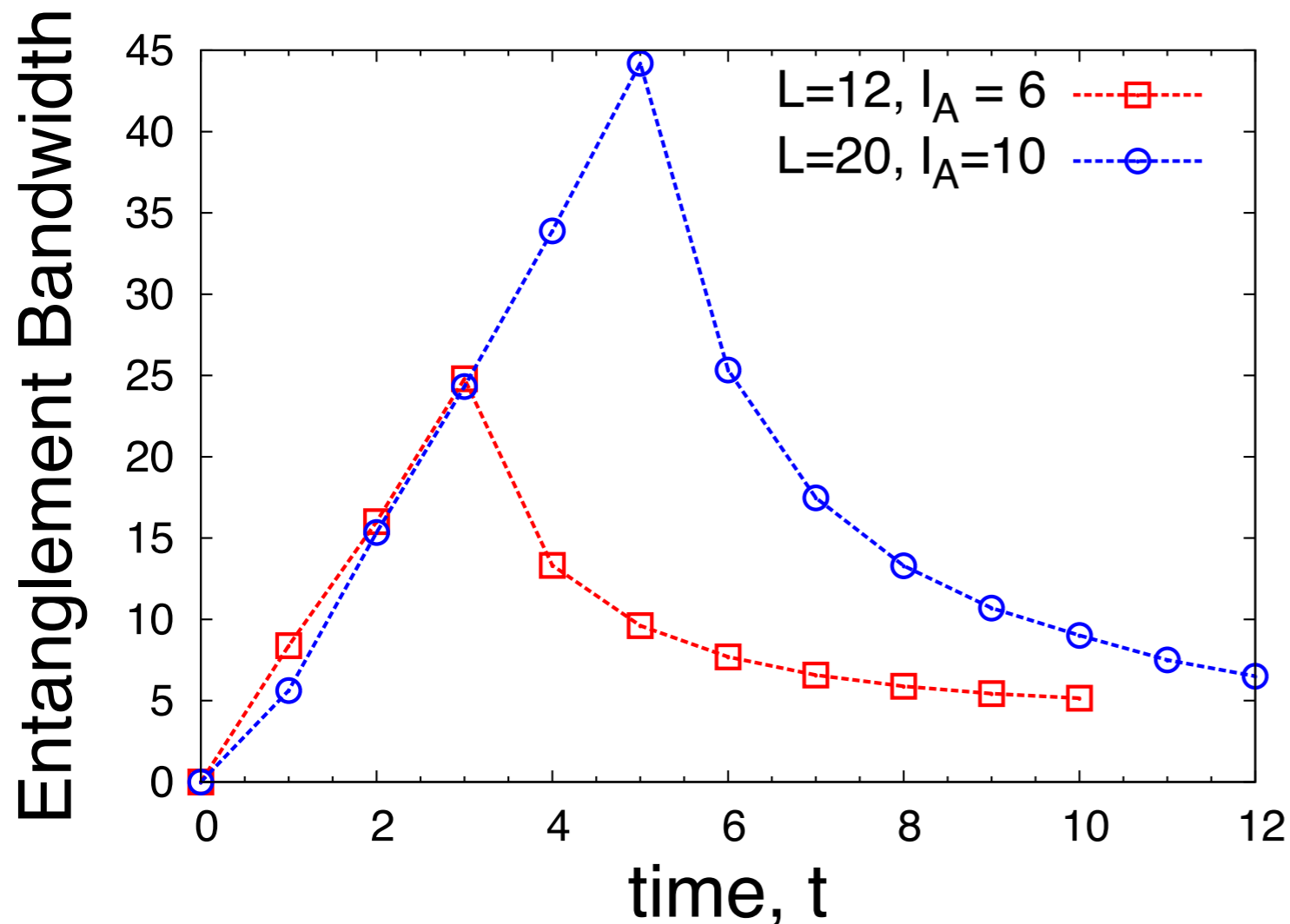
Growth of the rank of the reduced density matrix captured by the entanglement bandwidth!

Can be understood via operator spreading



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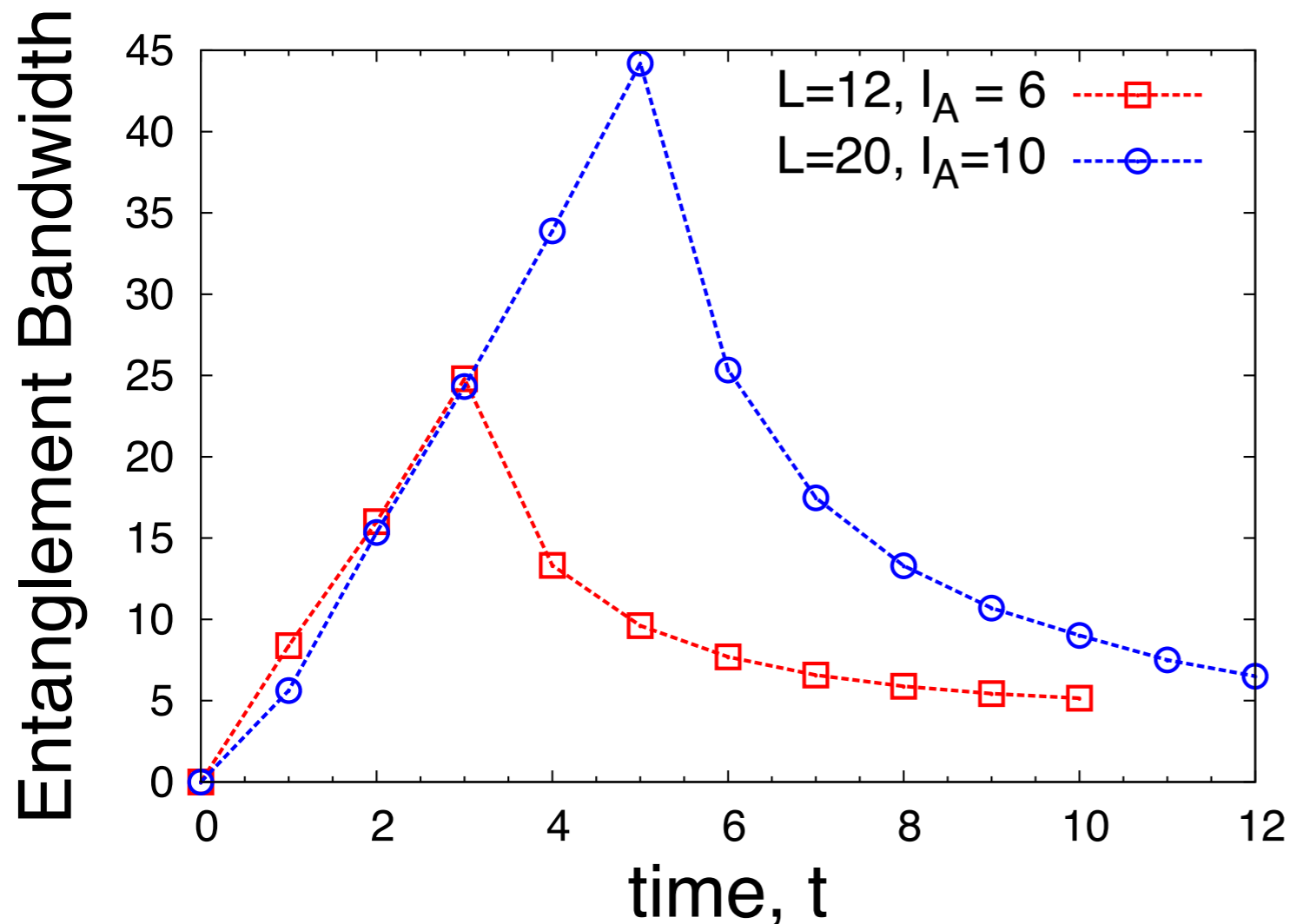
Operator carries weight

$$\sim \exp\{-[t(v_{LC} - v_B)]^2 / (Dt)\}$$

\sim RDM eigenvalues

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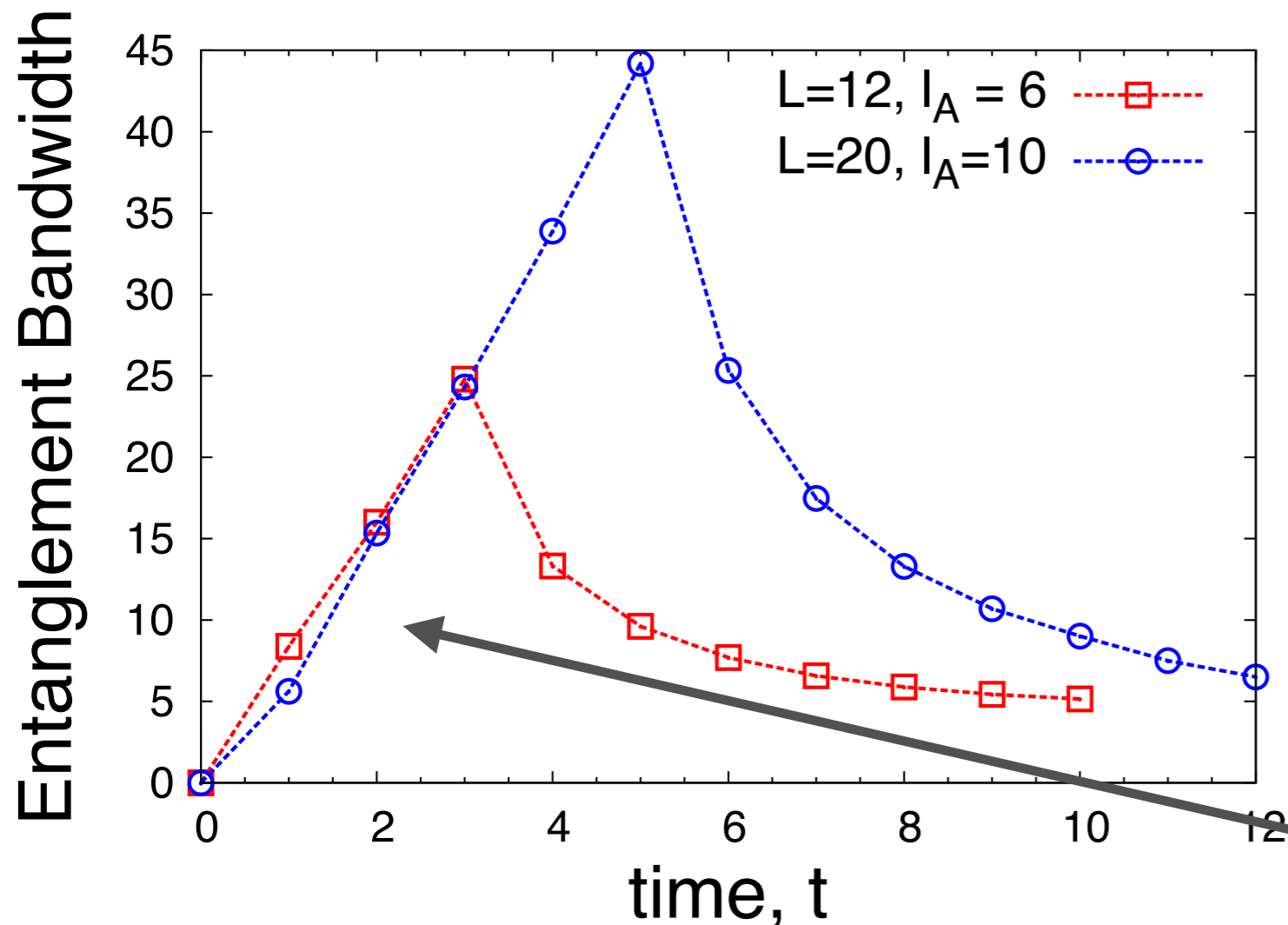
$$\sim \exp\{-[t(v_{LC} - v_B)]^2 / (Dt)\}$$

\sim RDM eigenvalues

$$\text{entanglement spectra} \sim \frac{(v_{LC} - v_B)^2}{D} t$$

ENTANGLEMENT ENTROPY VS BANDWIDTH

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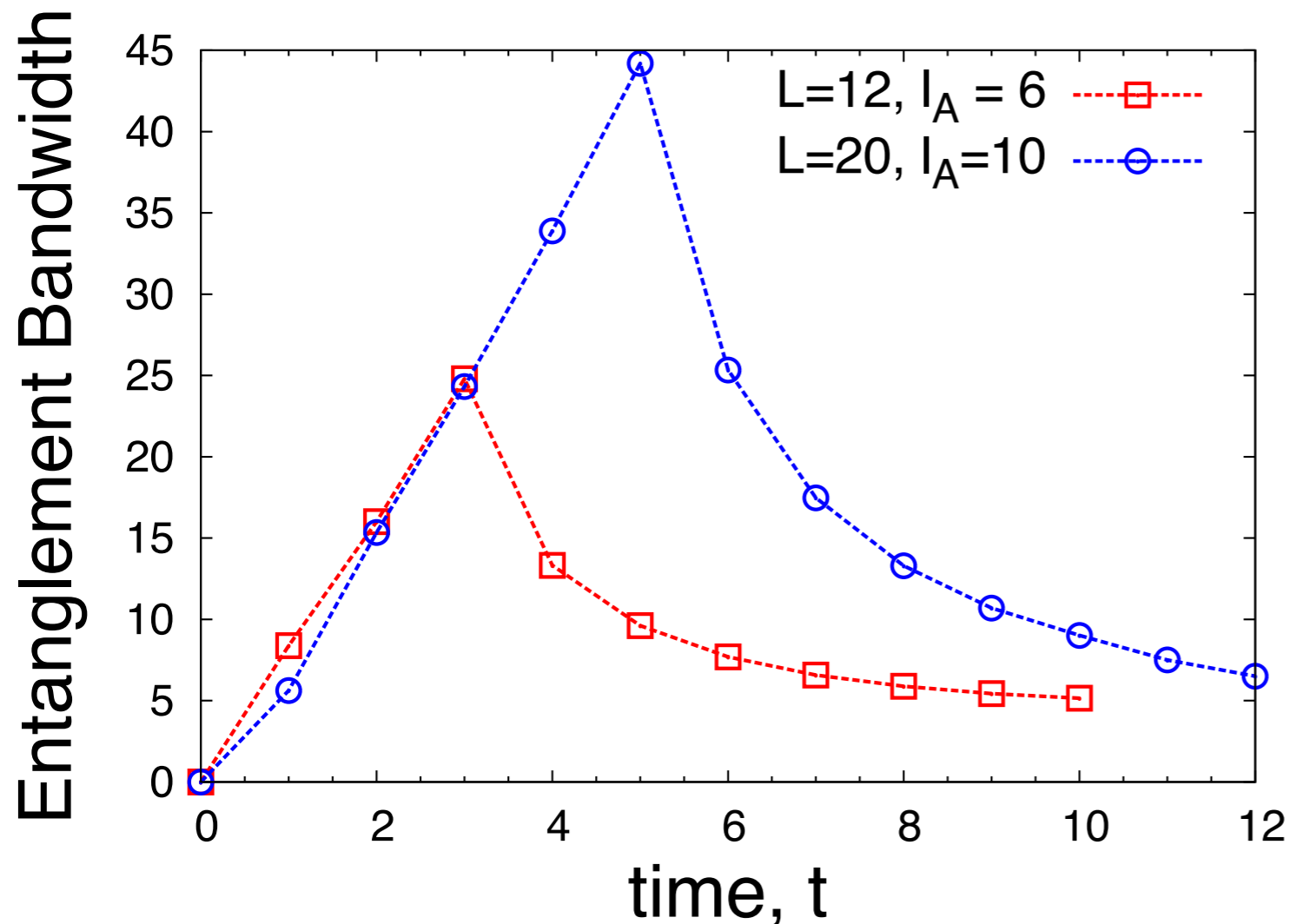
\sim RDM eigenvalues

entanglement spectra $\sim \frac{(v_{LC} - v_B)^2}{D} t$

bandwidth grows linearly!

ENTANGLEMENT ENTROPY VS BANDWIDTH

Growth of the rank of the reduced density matrix captured by the entanglement bandwidth!



Can be understood via
operator spreading

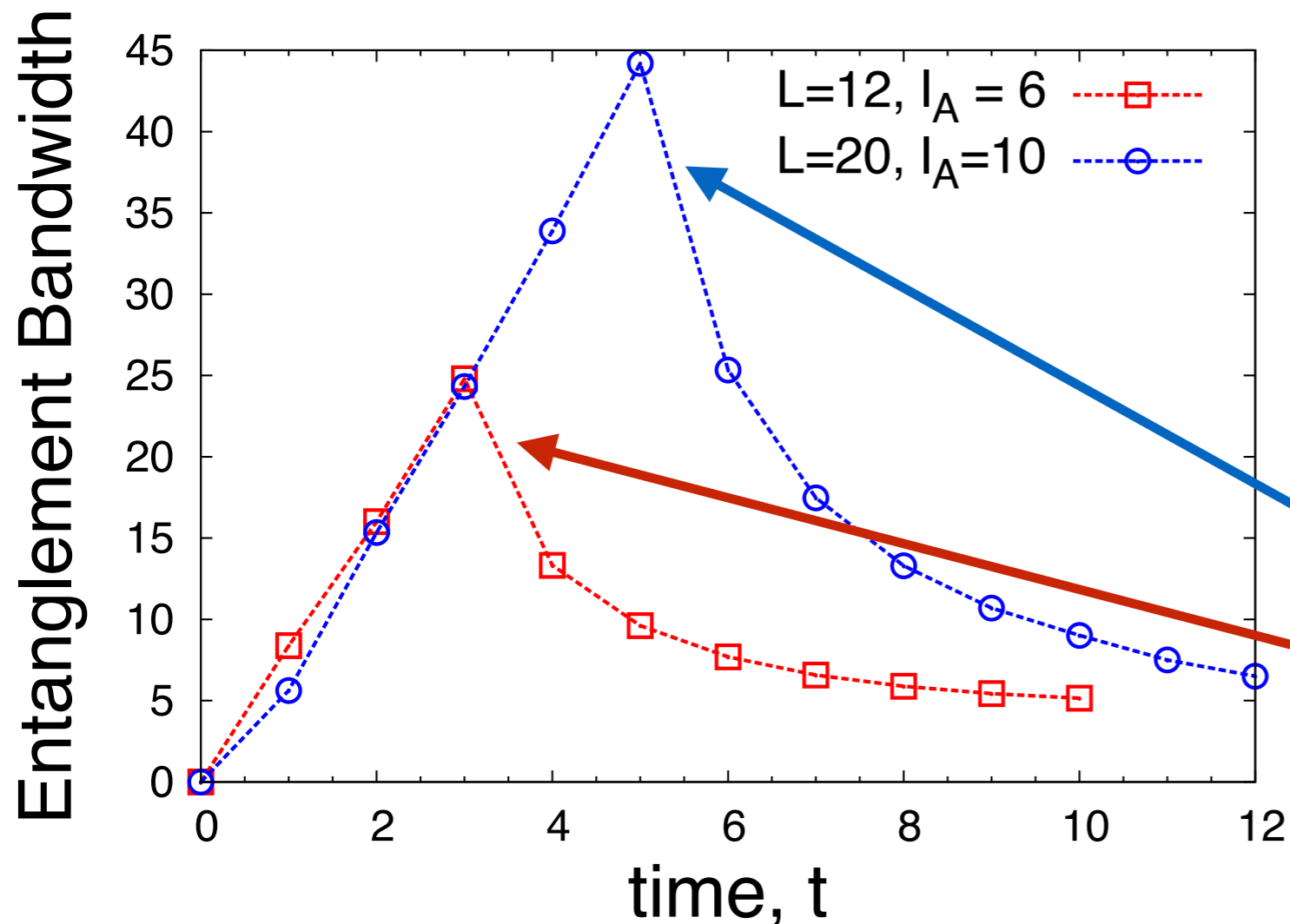
$$v_{LC} = \Delta x / \Delta t = 2$$

Light cone hits the edge of
the subsystem at time

$$t = l_A / v_{LC} = l_A / 2$$

ENTANGLEMENT ENTROPY VS BANDWIDTH

Growth of the rank of the reduced density matrix captured by the entanglement bandwidth!



Can be understood via
operator spreading

$$v_{LC} = \Delta x / \Delta t = 2$$

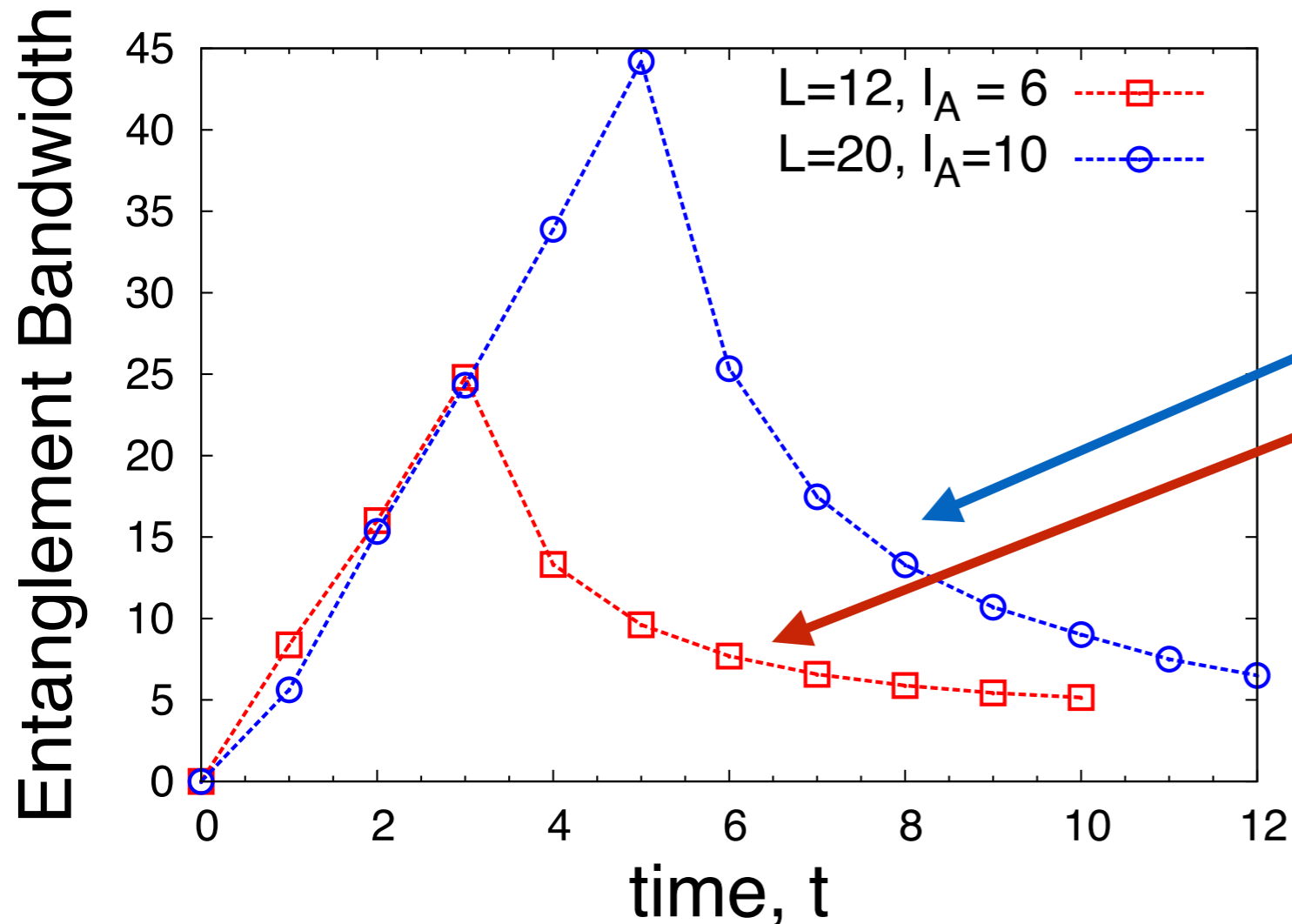
Light cone hits the edge of
the subsystem at time

$$t = l_A / v_{LC} = l_A / 2$$

Entanglement spectra's
bandwidth grows until
the light cone hits
the edge of the subsystem

ENTANGLEMENT ENTROPY VS BANDWIDTH

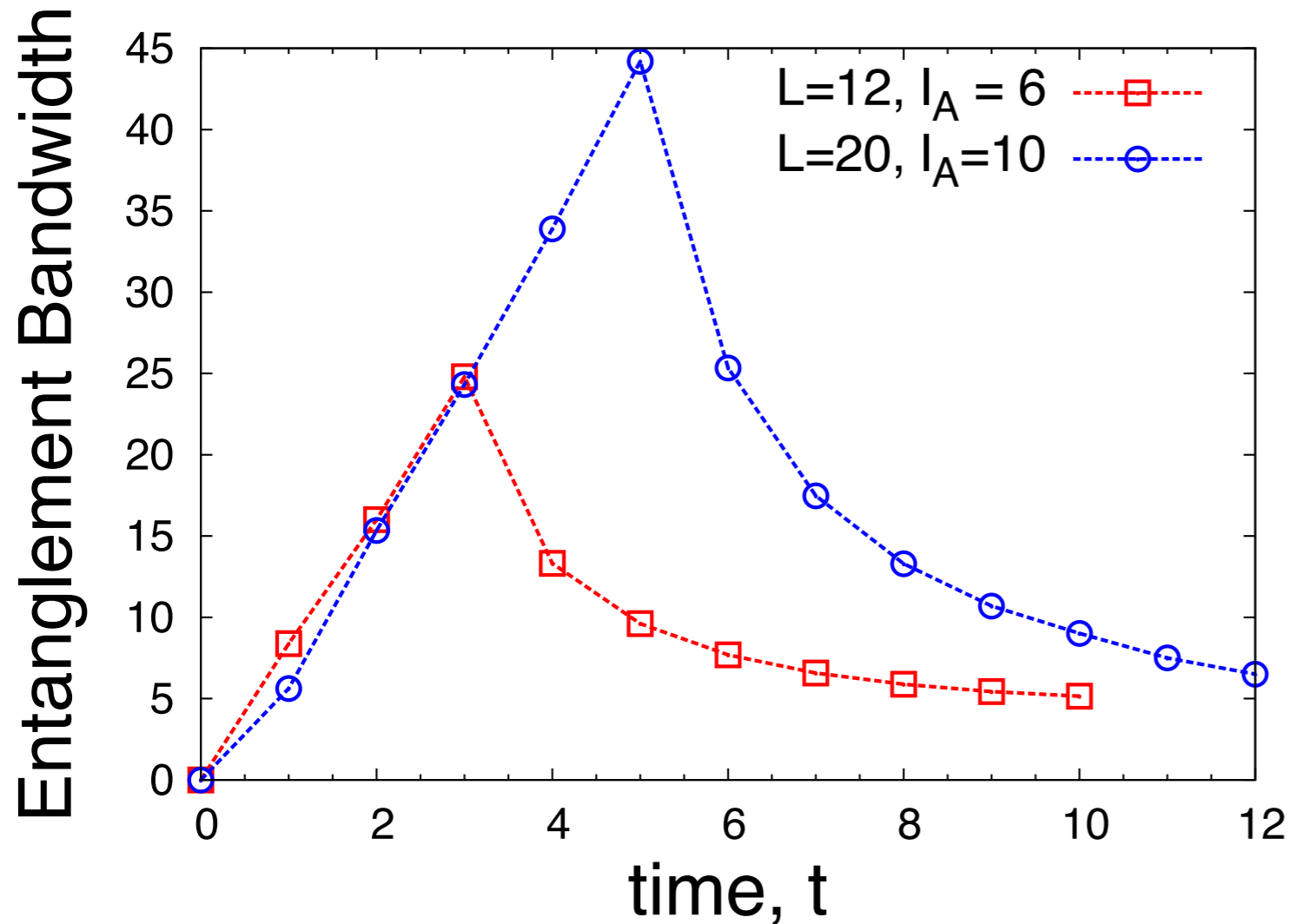
Growth of the rank of the reduced density matrix captured by the entanglement bandwidth!



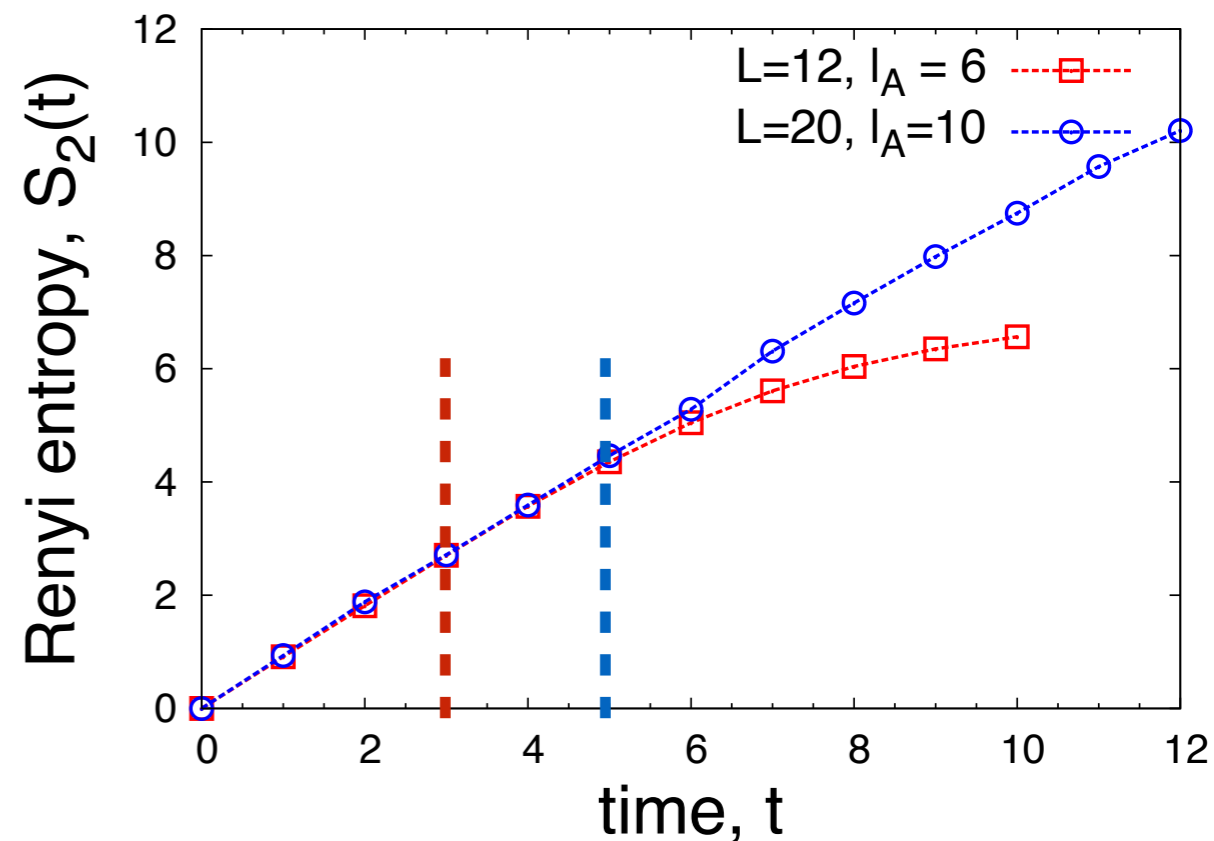
For later times
the RDM is full rank
 $= \min(q^{2t}, q^{l_A})$
and the BW shrinks like
 $\sim 1/(t - t_0)$
saturating to a value
 $\sim 1/\sqrt{q^{l_A}}$

ENTANGLEMENT ENTROPY VS BANDWIDTH

Growth of the rank of the reduced density matrix captured by the entanglement bandwidth!



This time scale is **invisible** in the entanglement entropy



SCRAMBLING AT SHORT TIMES, UNIVERSALITY?

At these short times

$$t < l_A/v_{LC}$$

is the distribution of
reduced density matrix
eigenvalues universal?

Consider the **distribution of
reduced density matrix
eigenvalues**

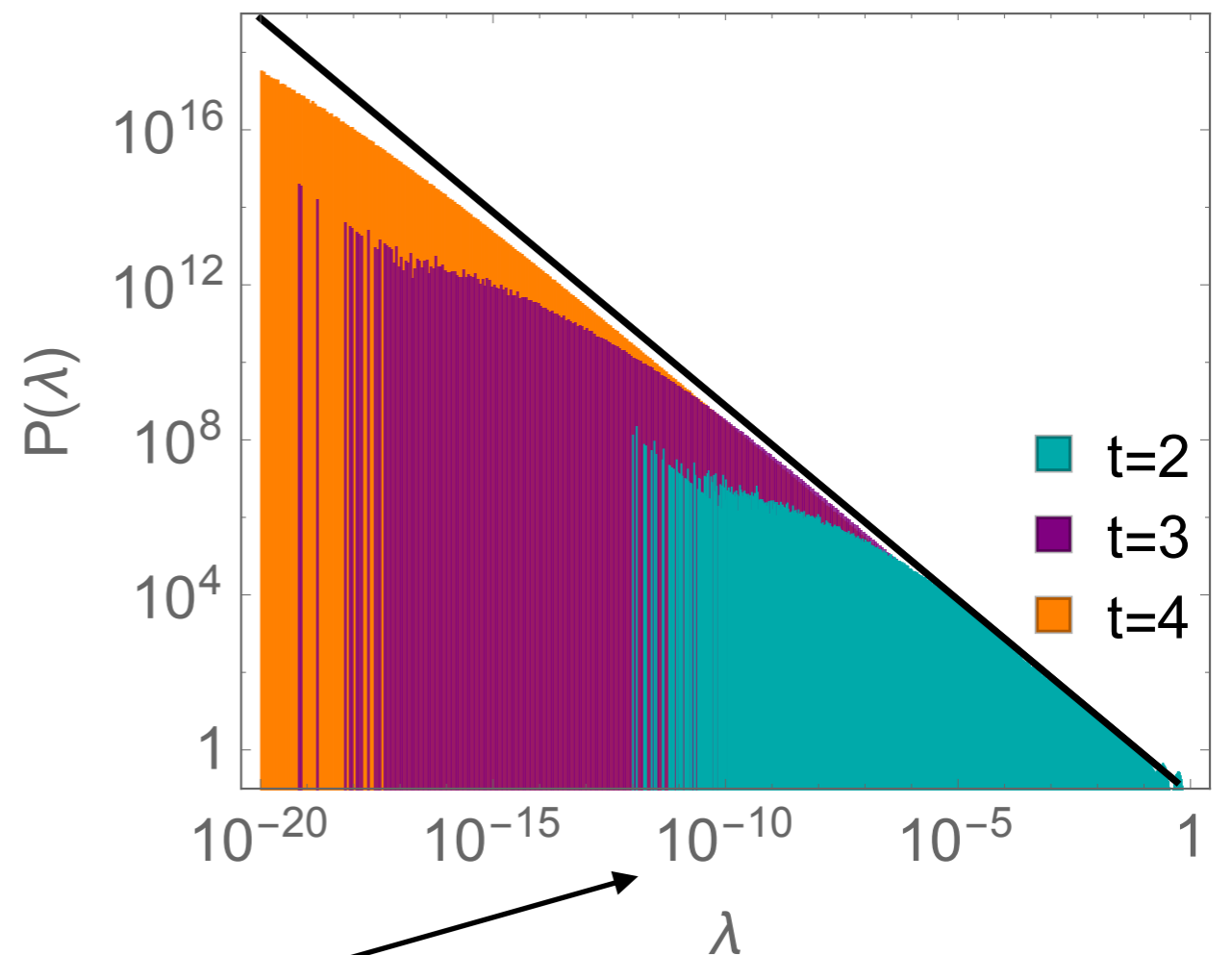
SCRAMBLING AT SHORT TIMES, UNIVERSALITY?

At these short times
 $t < l_A/v_{LC}$
is the distribution of
reduced density matrix
eigenvalues universal?

We find YES!

Consider the **distribution of
reduced density matrix
eigenvalues**

Reduced density matrix DOS



several decades of a scale free RDM eigenvalue distribution $P(\lambda) \sim 1/\lambda$

Demonstrated this for Hamiltonian and Floquet dynamics as well

SCRAMBLING AT SHORT TIMES, UNIVERSALITY?

At these short times
 $t < l_A/v_{LC}$
is the distribution of
reduced density matrix
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We find YES!

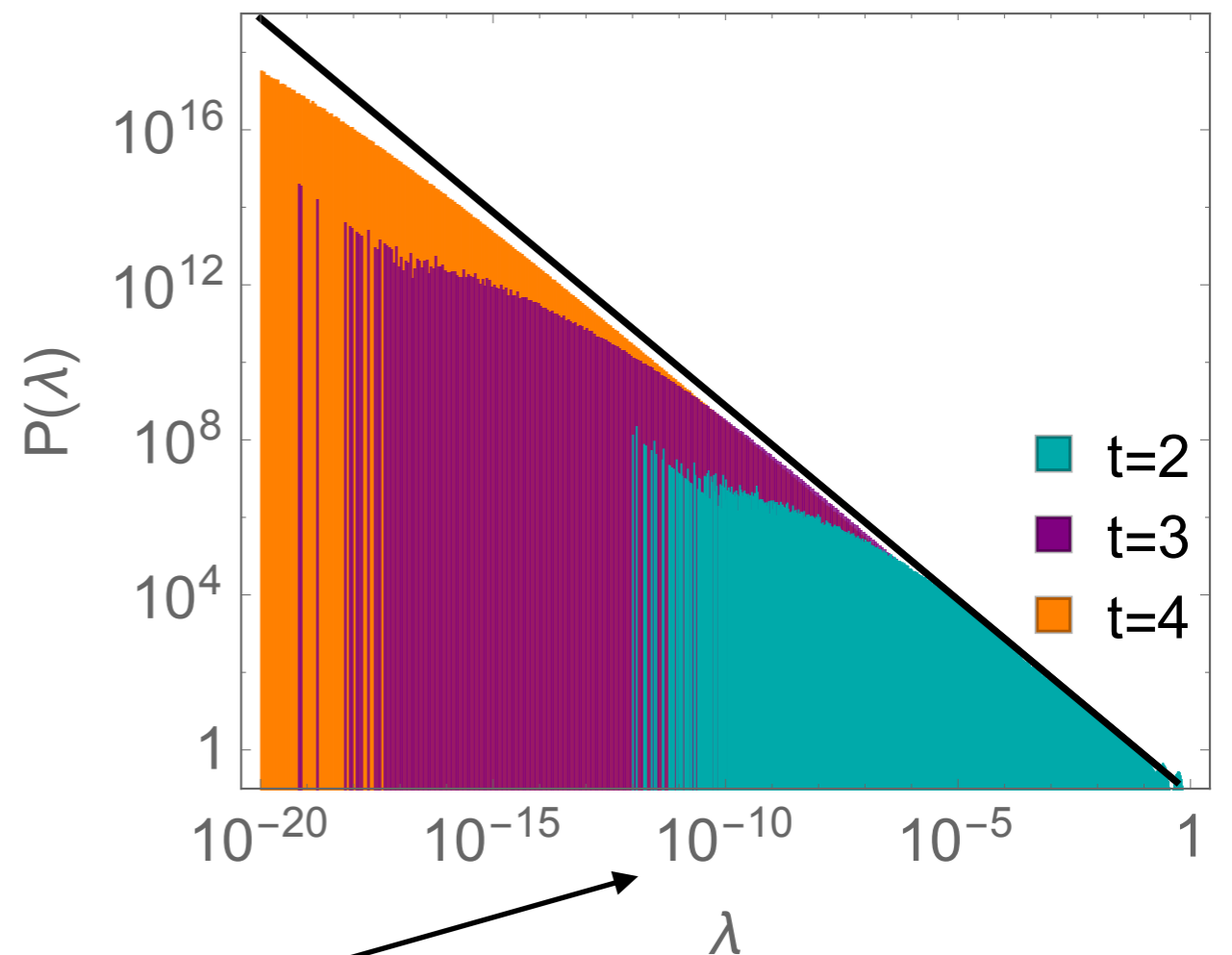
Consider the **distribution of
reduced density matrix
eigenvalues**

several decades of a scale free RDM eigenvalue distribution $P(\lambda) \sim 1/\lambda$

Distinct from a random thermal state
follows Marchenko-Pastur

$$P_{MP}(\lambda) \sim 1/\sqrt{\lambda}$$

Reduced density matrix DOS



SUMMARY: TIME SCALES ON APPROACH TO EQUILIBRIUM

Relevant time scales:

- (1) An $O(1)$ time scale, capturing local thermalization
- (2) When the light cone has hit the edge of the sub-system, $t \sim l_A$
 $0 < t < l_A/v_{LC}$
 $P(\lambda) \sim 1/\lambda$
- (3) Global thermalization time, when the entanglement has saturated $t \sim L$
 $P_{MP}(\lambda) \sim 1/\sqrt{\lambda}$

OUTLINE

- I. Lecture series layout
- II. Classical chaos
- III. Quantum chaos
- IV. Quantum thermalization
- V. Evading thermalization

HOW TO STOP SYSTEMS FROM THERMALIZING?

To stop a quantum system from thermalizing we need to
remove / reduce the entanglement

HOW TO STOP SYSTEMS FROM THERMALIZING?

To stop a quantum system from thermalizing we need to remove/reduce the entanglement

Many body localization (MBL): Strongly disordered or quasiperiodic many body Hamiltonians or Floquet unitaries

$$H = \sum_{i=1}^L [h_i \hat{S}_i^z + J \hat{\vec{S}}_i \cdot \hat{\vec{S}}_{i+1}]$$

Pal and Huse (2017)

$$\begin{aligned} H^{\text{QP/R}} = & J \sum_{i=1}^{L-1} (S_i^x S_{i+1}^x + S_i^y S_{i+1}^y) + J_z \sum_{i=1}^{L-1} S_i^z S_{i+1}^z \\ & + \sum_{i=1}^L W \cos(2\pi k i + \phi_i^{\text{QP/R}}) S_i^z \\ & + J' \sum_{i=1}^{L-2} (S_i^x S_{i+2}^x + S_i^y S_{i+2}^y), \end{aligned}$$

Khemani, Sheng, Huse (2017)

Setiawan, Deng, Pixley (2017)

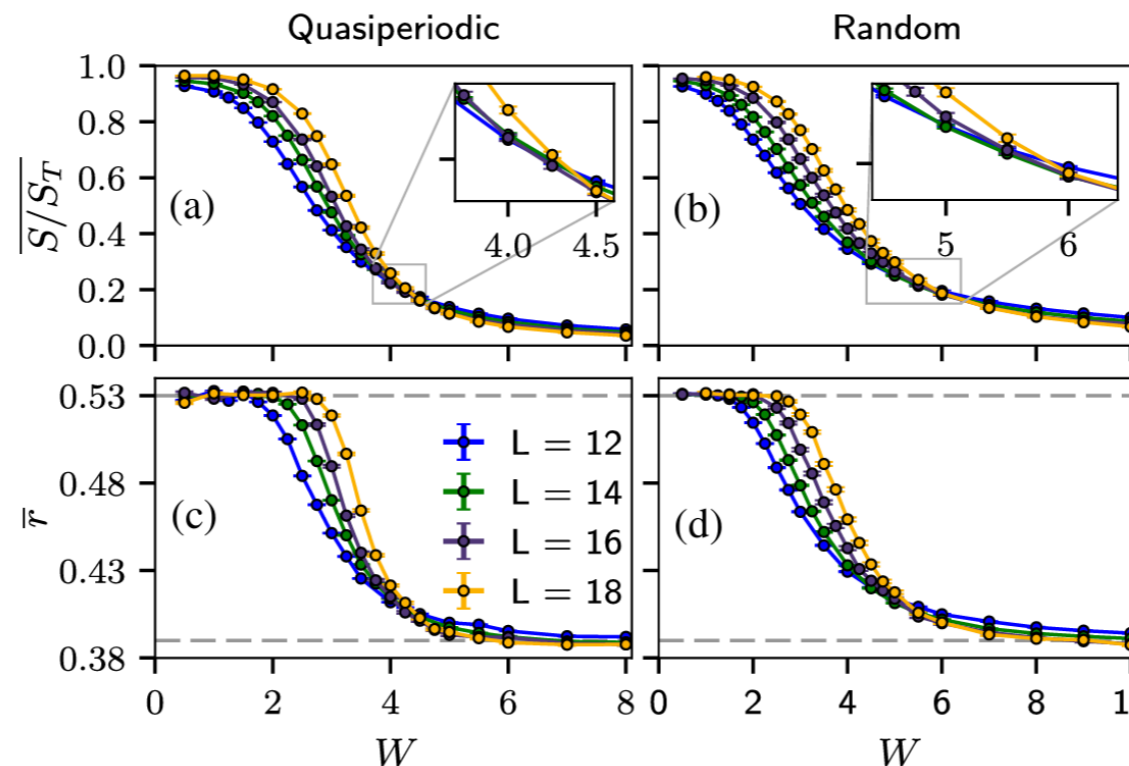
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To stop a quantum system from thermalizing we need to remove/reduce the entanglement

Many body localization (MBL): Strongly disordered or quasiperiodic many body Hamiltonians or Floquet unitaries

Spectral statistics on small sizes have found clear indication of either a transition or cross over in the entanglement and between random matrix theory and Poisson level statistics

Oganesyan and Huse, PRB (2007)



Khemani, Sheng, Huse (2017)

$$\delta_n = E_{n+1} - E_n$$

$$r_n = \min\{\delta_n, \delta_{n-1}\} / \max\{\delta_n, \delta_{n-1}\}$$

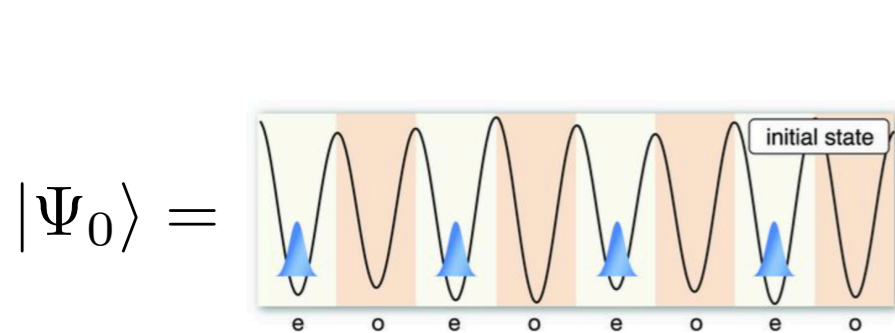
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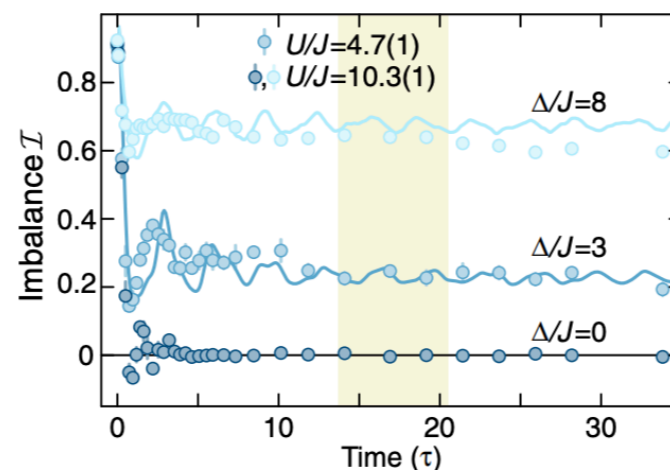
Many body localization (MBL): Strongly disordered or quasiperiodic many body Hamiltonians or Floquet unitaries

Experimental data on ultra cold atoms and trapped ions have found signatures of a transition, but are limited to finite time.

(Will focus on this in some detail in the next lecture, Friday June 5th)



$$|\Psi(t)\rangle = e^{-iHt}|\Psi_0\rangle$$



Schrieber et al, Science (2015)

Choi, et al Science (2016)

Smith, et al Nature Phys (2016)

Bordia, et al PRX (2017)

Guo, et al Nature Phys (2021)

HOW TO STOP SYSTEMS FROM THERMALIZING?

To stop a quantum system from thermalizing we need to
remove / reduce the entanglement

Many body localization (MBL): Strongly disordered or quasiperiodic many body
Hamiltonians or Floquet unitaries

Problem still strongly debated. Has been argued that the
relaxation time is exponentially large but finite, where as
MBL has an infinite relaxation time.

Disordered MBL not stable in dimensions greater than one.

De Roeck and Huveneers, PRB (2017)

Proof of MBL in 1D relies on one tacit assumption, not clear if
it is an ironclad proof.

Imbrie, Jour. Stat. Phys. (2016)

Huse, Physics Viewpoint (2016)

HOW TO STOP SYSTEMS FROM THERMALIZING?

To stop a quantum system from thermalizing we need to remove/reduce the entanglement

Quantum scars: Non-integrable Hamiltonians with a spectrum generating algebra $([H, \hat{Q}^\dagger] - \omega \hat{Q}^\dagger)W = 0,$

Invariant under Q and is a subspace of the full Hilbert space

Starting from some eigenstate $|\psi_0\rangle$

$(\hat{Q}^\dagger)^n |\psi_0\rangle$ Generates scar states that are **sub-thermal**
With energy $E_0 + n\hbar\omega$

HOW TO STOP SYSTEMS FROM THERMALIZING?

To stop a quantum system from thermalizing we need to remove/reduce the entanglement

Quantum scars: Non-integrable Hamiltonians with a spectrum generating algebra $([H, \hat{Q}^\dagger] - \omega \hat{Q}^\dagger)W = 0$,

Known examples

Spectrum-generating algebra

AKLT model, PXP model, spin-1 XY magnets, extended Hubbard model

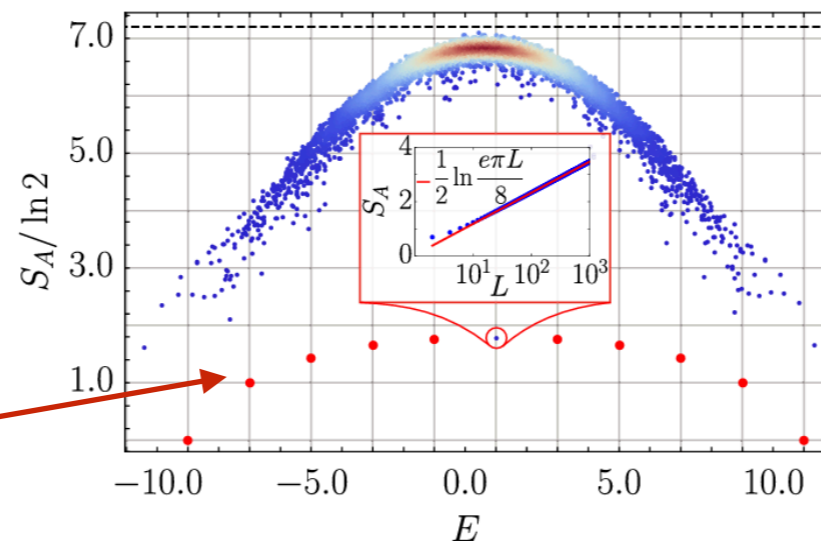
Scar states are log entangled subthermal

Has exact scar states

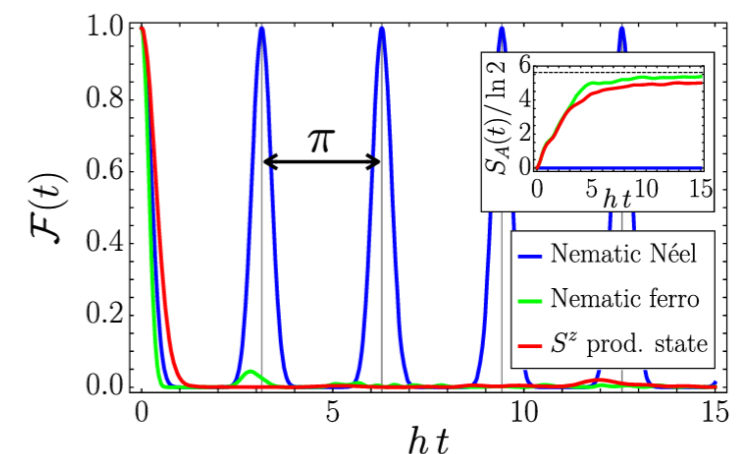
$$H = J H_{XY} + h S^z$$

$$= J \sum_r (S_r^x S_{r+1}^x + S_r^y S_{r+1}^y) + h \sum_r S_r^z$$

Schechter and Iadecola, PRL (2019)



Many-body revival of the Neel state



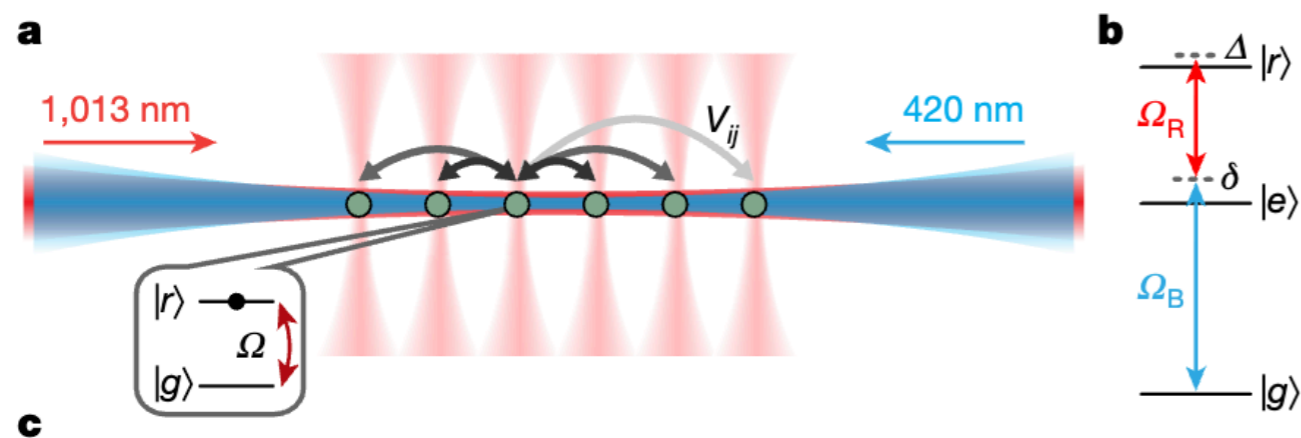
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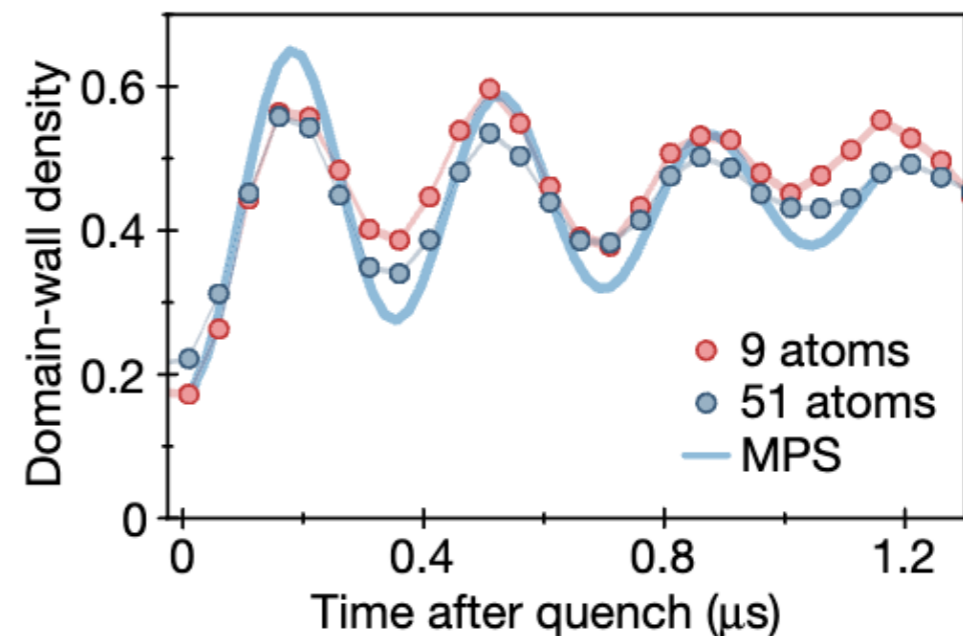
Effects of the scars seen in experiments observed anomalous relaxation in a Rydberg atom array

$$\frac{\mathcal{H}}{\hbar} = \sum_i \frac{\Omega_i}{2} \sigma_x^i - \sum_i \Delta_i n_i + \sum_{i<j} V_{ij} n_i n_j \quad \sigma_x^i = |g_i\rangle\langle r_i| + |r_i\rangle\langle g_i|$$

Rydberg atom tweezer arrays



Bernien et al Nature (2017)



Serbyn, Abanin, Papić (2021)

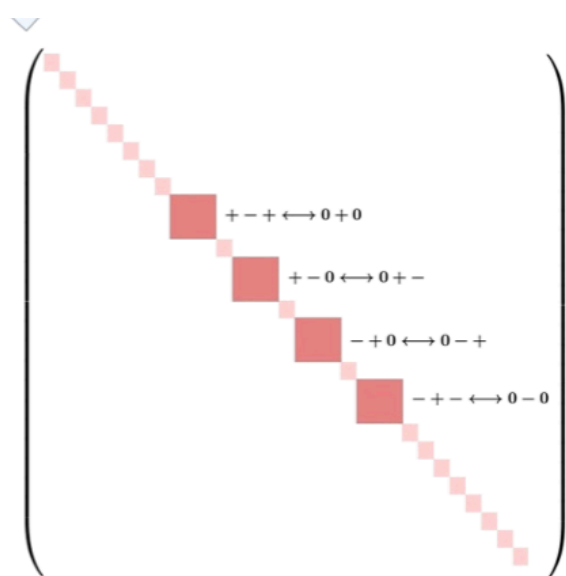
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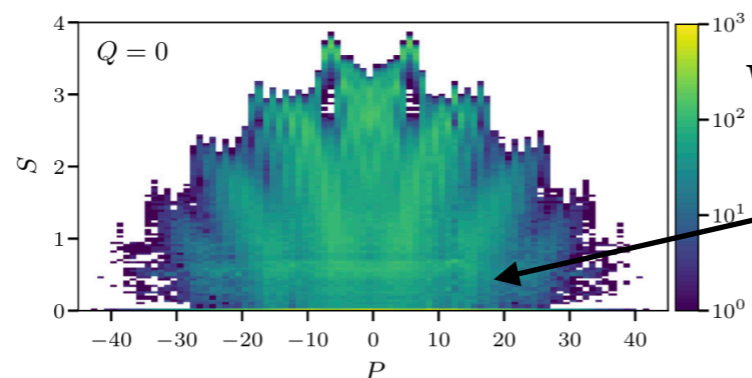
Hilbert space fragmentation: Constraints split the Hilbert space into many disconnected Krylov sectors. Prevents the dynamics from being ergodic.

Can be induced by a finite number of conservation laws

e.g. conserving charge and dipole

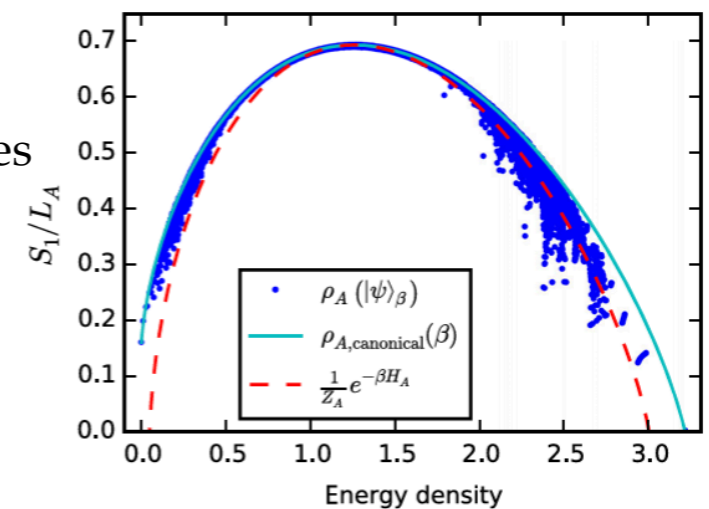


$$Q \equiv \sum_n S_n^z \quad \text{and} \quad P_{n_0} \equiv \sum_n (n - n_0) S_n^z$$



Weakly entangled states don't satisfy ETH

ETH



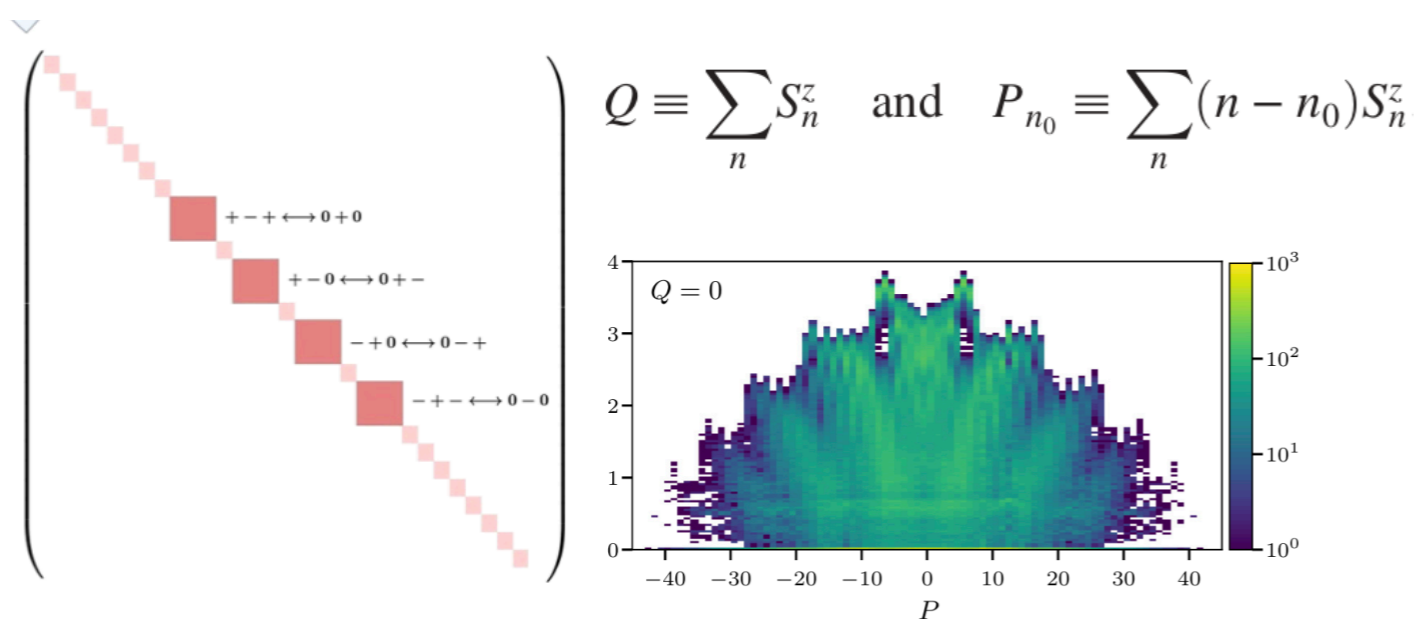
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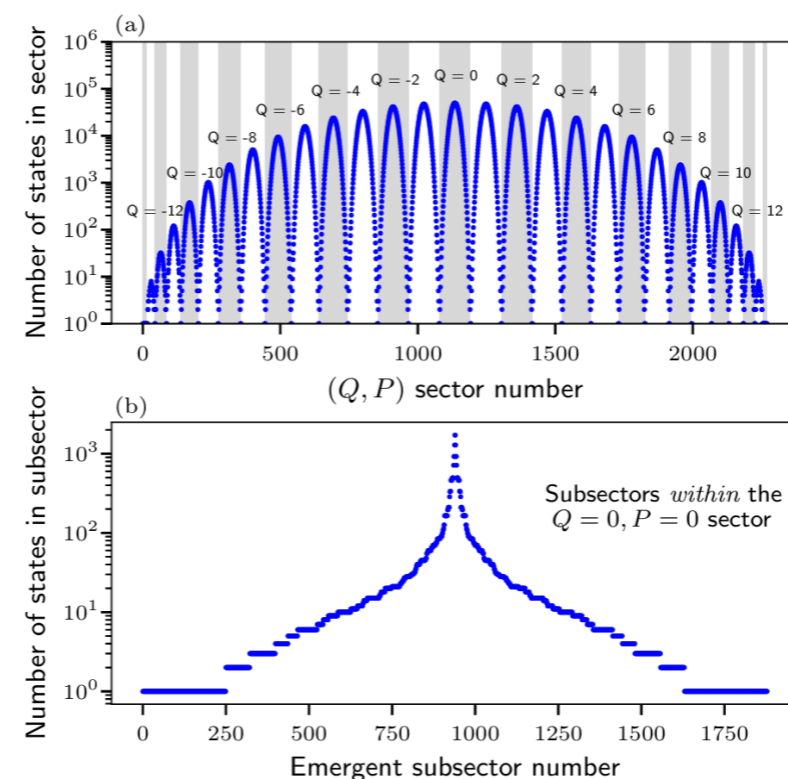
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HOW TO STOP SYSTEMS FROM THERMALIZING?

To stop a quantum system from thermalizing we need to
remove / reduce the entanglement

Measurement: Strong projective measurements collapse the
wave function and remove entanglement. By measuring a
particle's location we have localized it in a dynamical sense.

Focus of lecture 3 (June 8)

Feedback / feedforward: Based on the outcome of the
measurement perform an additional operation on the state

Focus of lecture 4 (June 10)

CONCLUSIONS

- Classical chaos is ubiquitous in nature.
- Quantum chaos of single particle systems remains well defined through random matrix theory.
- Many-body quantum chaos remains a challenging problem to define sharply as it lacks a simple classical limit.
- Thermalization in isolated many-body quantum systems takes place through locally scrambling information.
- Three relevant time scales on the approach to thermal equilibrium.
- Evading thermalization remains a topic of great interest.