

# On the relation between the magnitude and exponent of OTOCs

(Based on the paper with Alexei Kitaev [1812.00120])

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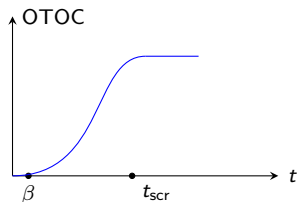
YITP, June 24, 2019

# Outline

- ▶ Introduction: OTOC in SYK
- ▶ Kinetic equation approach: an algorithm that computes the Lyapunov exponent
- ▶ Ladder identity and branching time
- ▶ Applications of the ladder identity

# Introduction

- ▶ OTOCs in large  $N$  systems with all-to-all interaction:



$$\text{Early time: } \langle A(t)B(0)A(t)B(0) \rangle_{\text{connected}} \sim \frac{1}{N} \frac{e^{\lambda_L t}}{C}$$

- ▶ Convention  $\beta = 2\pi$ . MSS chaos bound  $0 < \lambda_L \leq 1$ .
- ▶ Main example: SYK  $H_{\text{SYK}} = \sum_{jklm} J_{jklm} \chi_j \chi_k \chi_l \chi_m$

# OTOC in SYK

- ▶ At strong coupling  $J \gg 1$ , SYK is **almost** maximally chaotic (Maldacena-Stanford, 2016)

$$\text{OTOC}(t) \sim \frac{1}{N} \frac{e^{\lambda_L t}}{C}, \quad \lambda_L \sim 1 - \frac{c_1}{J}, \quad \frac{1}{C} \sim c_2 J$$

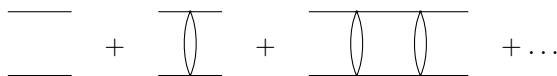
- ▶ Prefactor  $\frac{1}{C}$  is big: pseudo-Goldstone mode (Schwarzian mode).
- ▶ The ratio  $\frac{1-\lambda_L}{C}$  has a finite limit at  $J \rightarrow \infty$ .
- ▶ This talk is about an identity relating  $\lambda_L$  to  $C$ .  
[Corollary: although  $c_1, c_2$  individually depends on the UV data, the product  $c_1 c_2$  is universal.]
- ▶ Hope: improve the understanding of fast scrambling.

# Four point function and kinetic equation

Average over fermion indices, in general depends on four times.

$$\text{OTOC}(\underbrace{t_1, t_2}_{\text{future}}, \underbrace{t_3, t_4}_{\text{past}}) = \frac{1}{N^2} \sum_{jk} \langle \chi_j(t_1) \chi_k(t_3) \chi_j(t_2) \chi_k(t_4) \rangle + G(t_{12})G(t_{34})$$

- ▶ At large  $N$  limit, four-point functions are dominated by ladder diagrams (Kitaev, Polchinski-Rosenhaus, Maldacena-Stanford ...)

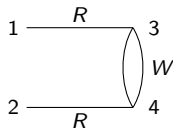


- ▶ Euclidean four-point function: Bethe-Salpeter equation

$$\mathcal{F} = \sum_n \mathcal{F}_n, \quad \mathcal{F}_n = \text{[Diagram: rectangle with a loop on the right side]} \cdot \mathcal{F}_{n-1} \Rightarrow \mathcal{F} = \mathcal{F}_0 + K\mathcal{F}$$

- ▶ OTOC: deform the contour to double Keldysh  $\Rightarrow$  Kinetic equation

$$\text{OTOC} \approx K^R \cdot \text{OTOC}, \quad K^R(t_1, t_2, t_3, t_4) :$$

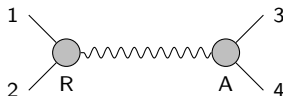


# Single-mode ansatz

How does  $\text{OTOC}(t_1, t_2, t_3, t_4)$  look like?  
*future* *past*

- ▶ We are interested in time regime  $t_1 \approx t_2 \gg t_3 \approx t_4$ : large separation between the perturbations in the past and probes in the future. (But still before the scrambling time.)
- ▶ Single-mode ansatz (“scramblon”) (Kitaev-Suh, 2017)

$$\text{OTOC}(t_1, t_2, t_3, t_4) \approx \frac{1}{N} \frac{e^{\lambda_L(t_1+t_2-t_3-t_4)/2}}{C} Y^R(t_{12}) Y^A(t_{34})$$



$Y^{R/A}$  : vertex function

## Eigenvalue $k_R(\alpha)$

The kinetic equation  $\text{OTOC} \approx K^R \text{OTOC}$  means OTOC is an eigenvector of  $K^R$  with eigenvalue  $k_R = 1$ . In general finding OTOC is a complicated search. We can reduce it to a shooting problem:

- ▶ Guess a growing exponent  $\alpha$  for scramblon  $e^{\alpha t}$ ;
- ▶ Adjust vertex function such that the trial  $\text{OTOC}(\alpha)$  is an eigenvector of  $K^R$  with general eigenvalue  $k_R(\alpha)$ ;
- ▶ Find  $\alpha_*$  such that  $k_R(\alpha_*) = 1 \Rightarrow \lambda_L = \alpha_*$ .

# Prefactor

This algorithm is useful in finding  $\lambda_L$  (as well as vertex functions), but the prefactor is ambiguous

- ▶ Because the kinetic equation  $\text{OTOC} \approx K^R \text{OTOC}$  is a homogeneous linear equation
- ▶ Origin of the ambiguity: we dropped the constant term  $F_0$  in the exact equation  $\text{OTOC} = K^R \text{OTOC} + F_0$
- ▶ We would like to get the correct prefactor with the data  $(\lambda_L, Y^{R/A}, k_R(\alpha))$  we already have



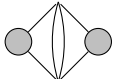
# Ladder identity

$$\text{OTOC}(t_1, t_2, t_3, t_4) \approx \frac{1}{N} \frac{e^{\lambda_L(t_1+t_2-t_3-t_4)/2}}{C} Y^R(t_{12}) Y^A(t_{34})$$

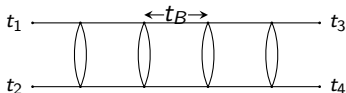
- ▶ Ladder identity:

$$\frac{2 \cos \frac{\lambda_L \pi}{2}}{C} \cdot (-k'_R(\lambda_L)) \cdot (Y^A, Y^R) = 1.$$

- ▶  $(Y^A, Y^R)$ : inner product of vertex functions:

$$(Y^A, Y^R) = \text{Diagram} = (q-1)J^2 \int dt Y^A(t) (G^W(t))^{q-2} Y^R(t).$$


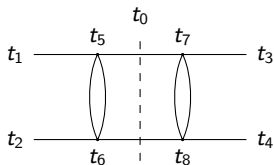
- ▶ Branching time  $t_B := -k'_R(\lambda_L)$ . Average separation between adjacent rungs.



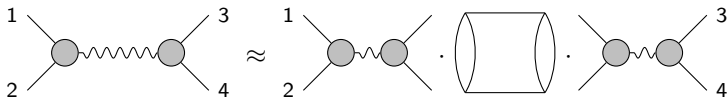
# Idea of the derivation

Idea: cut a long ladder into pieces and find a consistency condition.

- ▶ Cut in the middle, find adjacent  $\frac{t_5+t_6}{2} < t_0 < \frac{t_7+t_8}{2}$



- ▶ Cut-gluing consistency condition



- ▶ Left hand side contains one copy of  $C^{-1}$ , right hand side has two copies of  $C^{-1}$ .

# Applications

The ladder identity:

$$\frac{2 \cos \frac{\lambda_L \pi}{2}}{C} \cdot t_B \cdot (Y^A, Y^R) = 1.$$

Applications:

- ▶ Computational shortcuts:  $C \Leftrightarrow \lambda_L$ ;
- ▶ In a 1D model, exact maximal chaos.

# Computational shortcuts

- ▶  $C \Rightarrow \lambda_L$ : for SYK at strong coupling, the prefactor  $C$  is determined by the coefficient of the Schwarzian action  $\alpha_S$ . However, the Lyapunov exponent will receive small correction from the conformal matters (Maldacena-Stanford, 2016)

$$\delta\lambda_L \approx \frac{6\alpha_S}{Jk'_R(-1)\Delta(1-\Delta)(1-2\Delta)\tan(\pi\Delta)}.$$

- ▶  $\lambda_L \Rightarrow C$ : for  $q$ -body interacting SYK model at  $q \rightarrow \infty$  limit, the exact value of  $\lambda_L$  is easy to find using the kinetic equation (Maldacena-Stanford, 2016),

$$\frac{\lambda_L}{2 \cos \frac{\pi\lambda_L}{2}} = \mathcal{J}.$$

But the prefactor is hard to compute. The ladder identity gives a result consistent with the computation using Liouville equation (Qi-Streicher, 2018).

$$\text{OTOC}(t_1, t_2; t_3, t_4) \approx \frac{1}{N \cos \frac{\lambda_L \pi}{2}} \frac{e^{\lambda_L(t_1+t_2-t_3-t_4)/2}}{\left(2 \cosh \frac{\lambda_L t_{12}}{2}\right) \left(2 \cosh \frac{\lambda_L t_{34}}{2}\right)}.$$

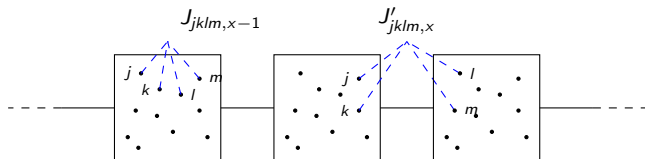
# Maximal chaos in a 1D model

- ▶ Idea: regard  $\lambda_L$  and  $C$  as analytic functions of some parameter,

$$\frac{2 \cos \frac{\lambda_L \pi}{2}}{C} \cdot t_B \cdot (Y^A, Y^R) = 1.$$

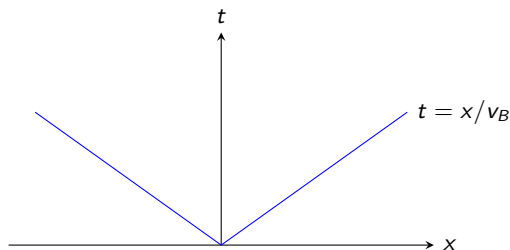
then the analytical properties of  $\lambda_L$  and  $C$  are locked by the ladder identity.

- ▶ A concrete example: SYK chain ([YG-Qi-Stanford, 2016](#))



# Operators at two different locations

Spatial propagation of chaos: measured by  $\langle A(x, t)B(0, 0)A(x, t)B(0, 0) \rangle$



- ▶ Fourier transform:

$$\text{OTOC}(x, t) = \int \frac{dp}{2\pi} e^{ipx} \text{OTOC}(p, t)$$

- ▶ Each  $\text{OTOC}(p, t)$  can be solved using retarded kernel approach.

# Fourier Transform

- ▶ The ladder identity holds for each  $\text{OTOC}(p, t)$ :

$$C(p) = 2 \cos \frac{\lambda_L(p)\pi}{2} \cdot t_B \cdot (Y^A, Y^R),$$

- ▶ The dependence of  $t_B$  and  $(Y^A, Y^R)$  on  $p$  is not important (analytic and do not vanish in the domain of interest).

$$\text{OTOC}(x, t) \sim \frac{1}{N} \int_{-\infty}^{+\infty} \underbrace{\frac{dp}{2\pi} \frac{e^{\lambda_L(p)t+ipx}}{2 \cos \frac{\pi\lambda_L(p)}{2}}}_{u(x,t)} \cdot \frac{Y^R Y^A}{t_B(Y^A, Y^R)}.$$

# Butterfly wavefront

$$u(x, t) = \int_{-\infty}^{+\infty} \frac{dp}{2\pi} \frac{e^{\lambda_L(p)t + ipx}}{2 \cos \frac{\lambda_L(p)}{2}}$$

- ▶ Butterfly wavefront:  $u(x, t) \sim 1$ .
- ▶ For large  $x > 0$  and  $t$ , we can estimate the asymptotics by saddle point of the exponent.

$$\lambda'_L(p)t + ix = 0, \quad p = i|p|.$$

- ▶ The relevant saddle point  $p_s = i|p_s|$  is purely imaginary: deform the integral contour to pass. **Warning: pole contribution!**

$$\cos \frac{\lambda_L(p_1)\pi}{2} = 0, \quad \lambda_L(p_1) = 1, \quad p_1 = i|p_1|.$$

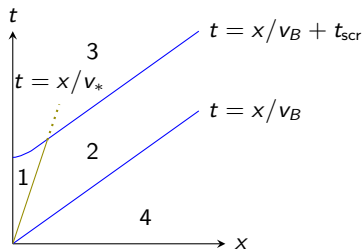


# Pole contribution: maximal chaos

- ▶ When  $J > J_c$ , the pole dominates the butterfly wavefront:

$$u_1(x, t) \approx \frac{e^{t - |p_1||x|}}{\pi i \lambda'_L(p_1)}, \quad v_1 = \frac{1}{|p_1|}$$

- ▶ Four regions with different OTOC behavior.



# Summary and discussion

An identity relates  $C$  and  $\lambda_L$ ;

$$\frac{2 \cos \frac{\lambda_L \pi}{2}}{C} \cdot t_B \cdot (Y^A, Y^R) = 1, \quad t_B = (-k'_R(\lambda_L))$$

Some lessons from the applications:

- ▶ Computational shortcuts: in SYK model we find the correction  $\delta\lambda_L \propto t_B^{-1}$ . If we call  $\lambda_L = 1$  the “coherent scrambling” (Schwarzian, gravity...), then the branching time  $t_B$  measures the “decoherence” effect.
- ▶ Maximal chaos: spatial locality provides a mechanism to enhance the Lyapunov exponent. Ladder identity: the enhancement cancels the correction exactly

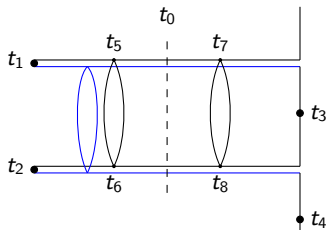
**Thank you!**

# Derivation: $\cos \frac{\lambda_L \pi}{2}$ factor

- ▶ Naively, we would have a formula:

$$\text{OTOC} \approx \text{OTOC}_L \cdot \text{BOX} \cdot \text{OTOC}_R$$

- ▶ Subtlety: multiple choices on the double Keldysh contour

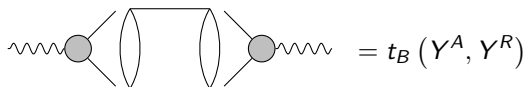


- ▶ Sum of two choices:

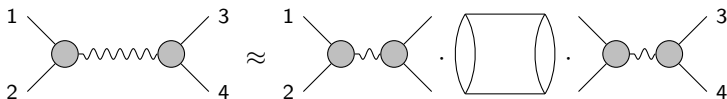
$$\text{OTOC} \approx \left( e^{i \frac{\lambda_L \pi}{2}} + e^{-i \frac{\lambda_L \pi}{2}} \right) \text{OTOC}_L \cdot \text{BOX} \cdot \text{OTOC}_R$$

# Derivation

- ▶ Gluing  $\text{OTOC}_L$  and  $\text{OTOC}_R$  through a box:


$$= t_B(Y^A, Y^R)$$

- ▶ Now inserting the single mode ansatz to the consistency condition  
 $\text{OTOC} \approx \left( e^{i\frac{\lambda_L\pi}{2}} + e^{-i\frac{\lambda_L\pi}{2}} \right) \text{OTOC}_L \cdot \text{BOX} \cdot \text{OTOC}_R$


$$\approx \text{[diagram of product of three terms]}$$

- ▶ Compare two sides (stripping off the vertex functions at two ends)

$$\frac{1}{C} = 2 \cos \frac{\lambda_L \pi}{2} \cdot \frac{1}{C} \cdot t_B \cdot (Y^A, Y^R) \cdot \frac{1}{C}$$