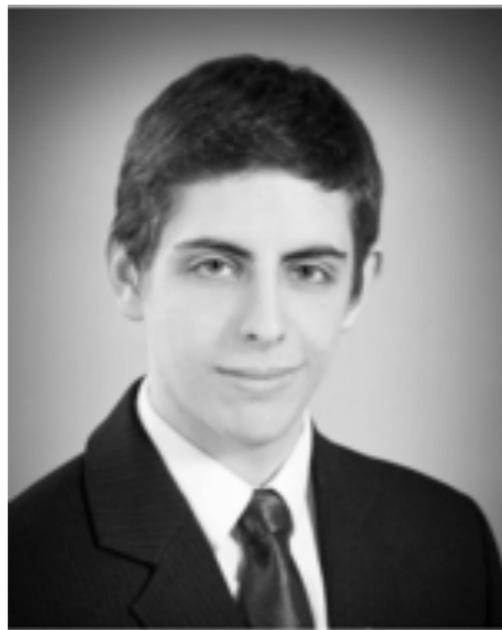


Efficient simulation of quantum transport in 1D

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T. Rakovszky, Stanford



C.v Keyserlingk, Birmingham

[Rakovszky, von Keyserlingk, FP, PRB 105, 07513 (2022)]

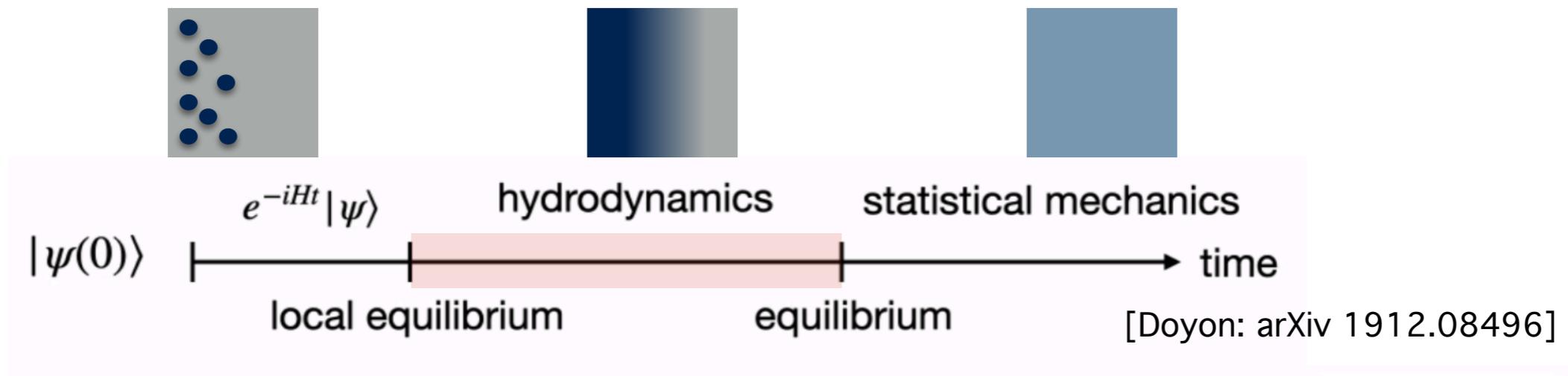
[von Keyserlingk, FP, Rakovszky PRB 105, 245101 (2022)]



Mathematical results of many-body
quantum systems
Herrsching, June 10, 2022

Characterizing thermalization dynamics

Universal hydrodynamic features tend to emerge in the low-frequency, long-wavelength limit

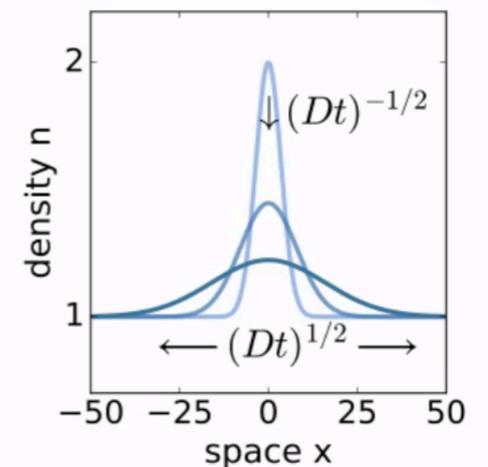


Emergent hydrodynamic relaxation:

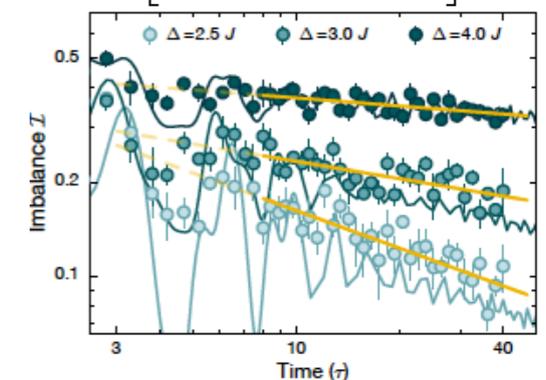
Diffusion

$$\partial_t n = -\partial_x j \quad j = -D\partial_x n \quad (\text{Fick's law})$$

$$\longrightarrow \partial_t n = D\partial_x^2 n \quad (\text{Diffusion})$$



[Lüschen et al. '17]



How to numerically extract hydrodynamics for a given microscopic model?

Numerical complexity of many-body dynamics

Directly simulate the time evolution within the full many-body Hilbert space $|\psi(t)\rangle = e^{-itH} |\psi(0)\rangle$

$$|\psi\rangle = \sum_{j_1, j_2, \dots, j_L} \psi_{j_1, j_2, \dots, j_L} |j_1\rangle |j_2\rangle \cdots |j_L\rangle, \quad j_n = 1 \dots d$$

- ▶ Complexity $\propto \exp(L)$
- ▶ Exact diagonalization methods
(dynamical typicality) up to ~ 30 spins

Matrix-Product States

Low entanglement: Matrix-Product States $d^L \rightarrow Ld\chi^2$

[M. Fannes et al. 92]

$$\psi_{j_1, j_2, \dots, j_L} \approx \sum_{\alpha_1, \alpha_2, \dots, \alpha_{L-1}} A_{\alpha_1}^{j_1} A_{\alpha_1, \alpha_2}^{j_2} \dots A_{\alpha_{L-1}}^{j_L} \quad (\alpha_j = 1 \dots \chi)$$

- ▶ Efficient representation of ground states of gapped local Hamiltonians [Hastings, Schuch, Verstraete, ...]

Diagrammatic representation

$$\psi_{j_1, j_2, j_3, j_4, j_5} \approx \begin{array}{c} | \\ \bullet \\ A^{[1]} \end{array} \begin{array}{c} | \\ \bullet \\ A^{[2]} \end{array} \begin{array}{c} | \\ \bullet \\ A^{[3]} \end{array} \begin{array}{c} | \\ \bullet \\ A^{[4]} \end{array} \begin{array}{c} | \\ \bullet \\ A^{[5]} \end{array}$$

$$A_{\alpha, \beta}^j = \begin{array}{c} j \\ | \\ \alpha \text{---} \bullet \text{---} \beta \\ A \end{array}$$

$$\begin{aligned} \alpha, \beta &= 1 \dots \chi \\ j &= 1 \dots d \end{aligned}$$

Numerical complexity of many-body dynamics

Directly simulate the time evolution within the full many-body Hilbert space $|\psi(t)\rangle = e^{-itH}|\psi(0)\rangle$

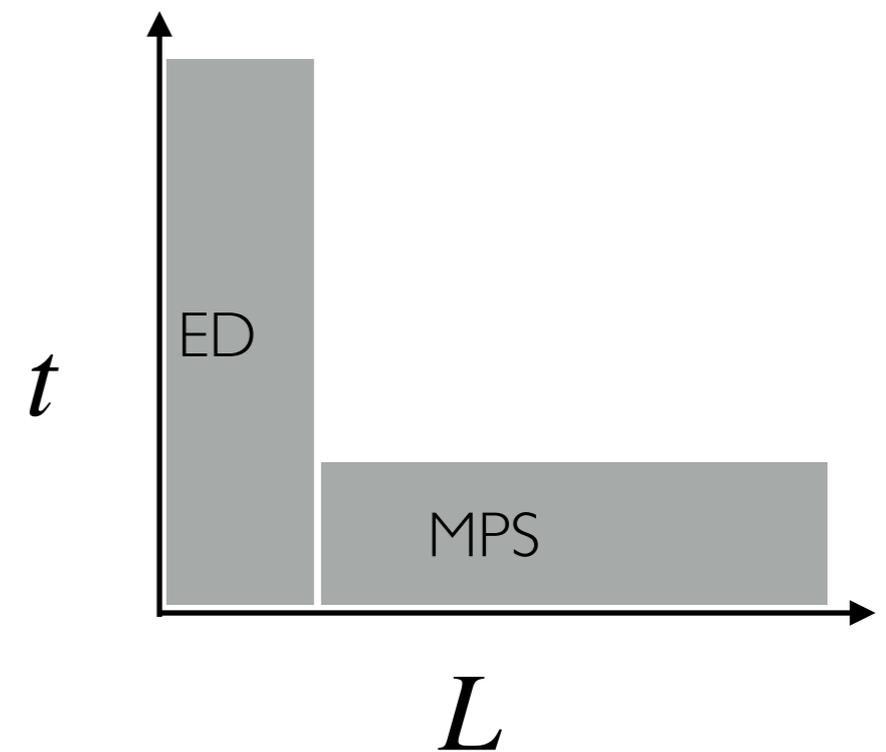
$$|\psi\rangle = \sum_{j_1, j_2, \dots, j_L} \psi_{j_1, j_2, \dots, j_L} |j_1\rangle |j_2\rangle \dots |j_L\rangle, \quad j_n = 1 \dots d$$

- ▶ Complexity $\propto \exp(L)$
- ▶ Sparse exact diagonalization methods (dynamical typicality) up to ~ 30 spins

Matrix-Product State based numerics

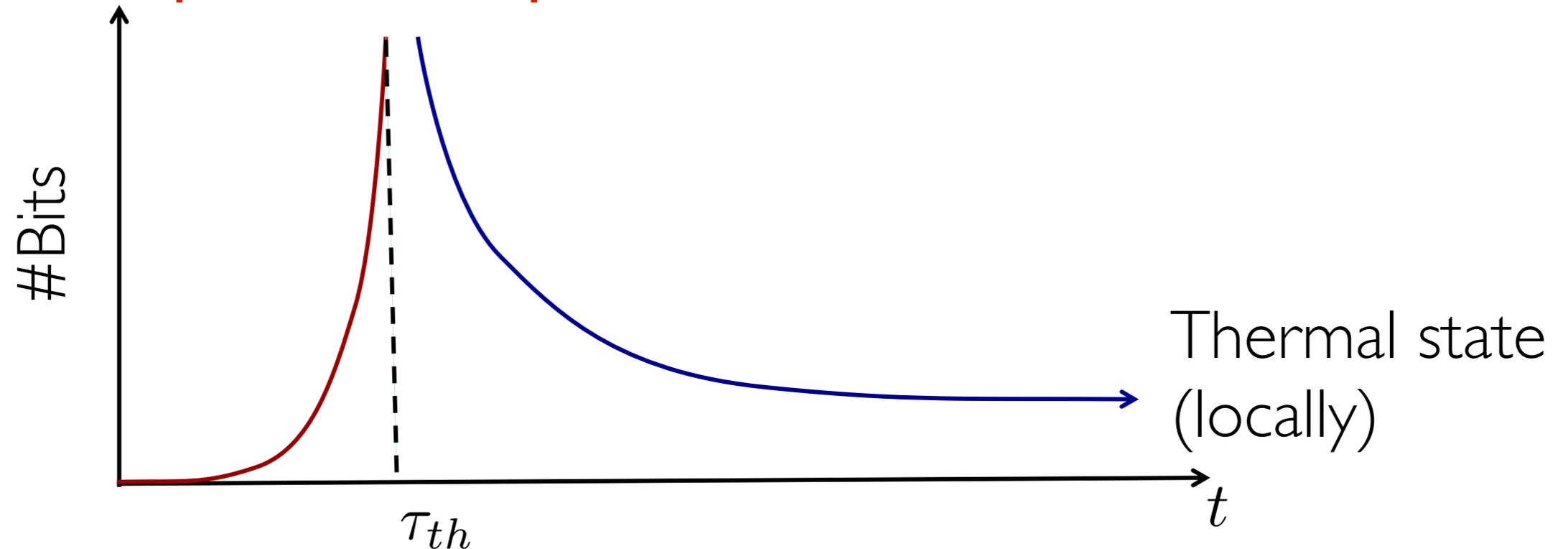
- ▶ Complexity $\propto \exp(t)$ because of linear entanglement growth

[P. Calabrese and J. Cardy, Huse, Nahum]



“Information paradox”

Quantum quench from product state



How to truncate entanglement without sacrificing crucial information on physical (local) observables?

Various approaches to address this problem:

[White et al.: PRB 2018]

[Wurtz et al.: Ann. Phys. 2018]

[Klein Kvorning, arXiv:2105.11206]

[Schmitt, Heyl: SciPost 2018]

[Parker et al., PRX 2019]

[Krumnow et al.: arXiv:1904.11999]

[Leviatan et al., arXiv:1702.08894]

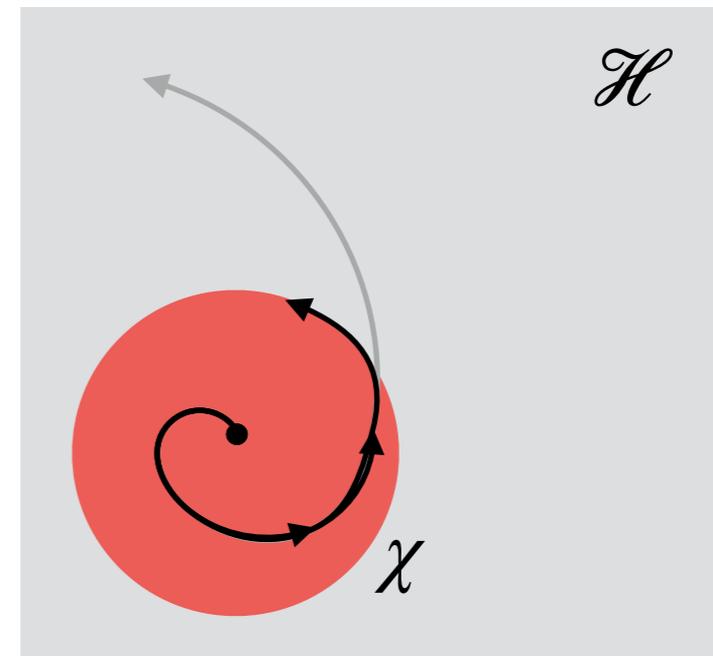
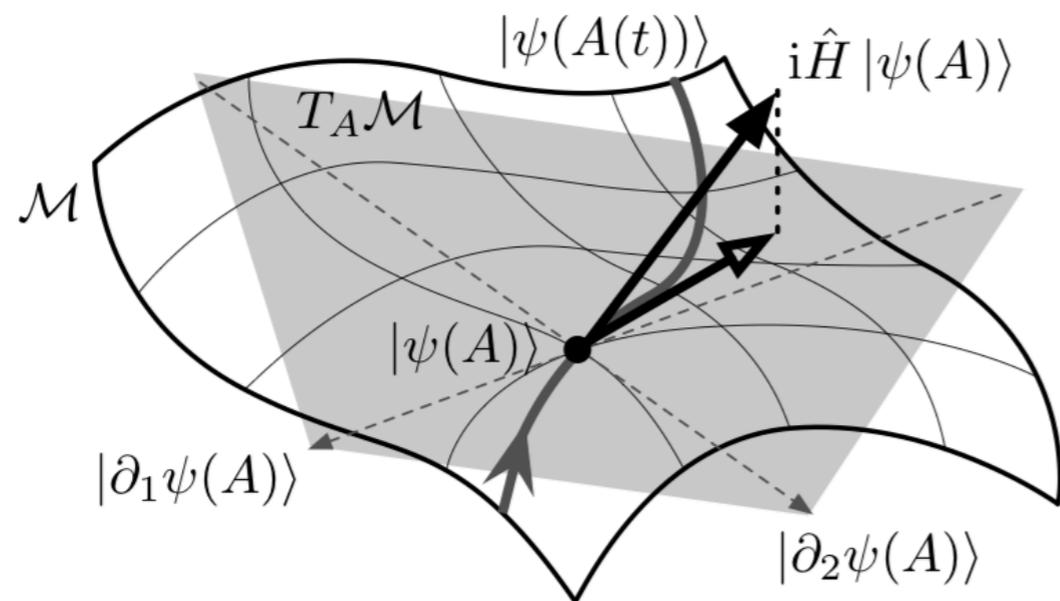
Time-dependent variational principle (TDVP)

Variational manifold: MPS states with fixed bond dimension

$$\psi_{j_1, j_2, j_3, j_4, j_5} = A_{\alpha}^{[1]j_1} A_{\alpha\beta}^{[2]j_2} A_{\beta\gamma}^{[3]j_3} A_{\gamma\delta}^{[4]j_4} A_{\delta}^{[5]j_5}$$

Classical Lagrangian $\mathcal{L}[\alpha, \dot{\alpha}] = \langle \psi[\alpha] | i\partial_t | \psi[\alpha] \rangle - \langle \psi[\alpha] | H | \psi[\alpha] \rangle$

Efficient evolution using a projected Hamiltonian [Haegeman et al. '11, Dorando et al. '09]

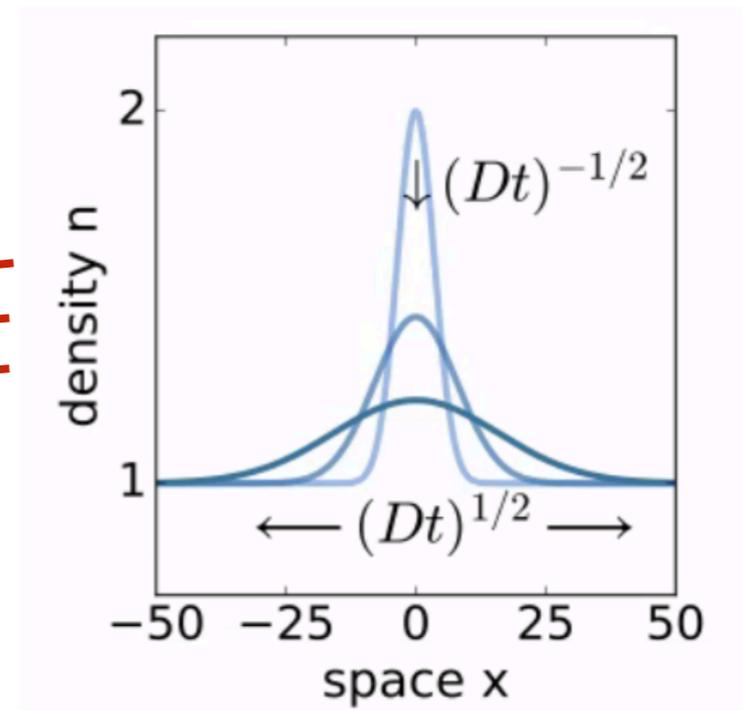
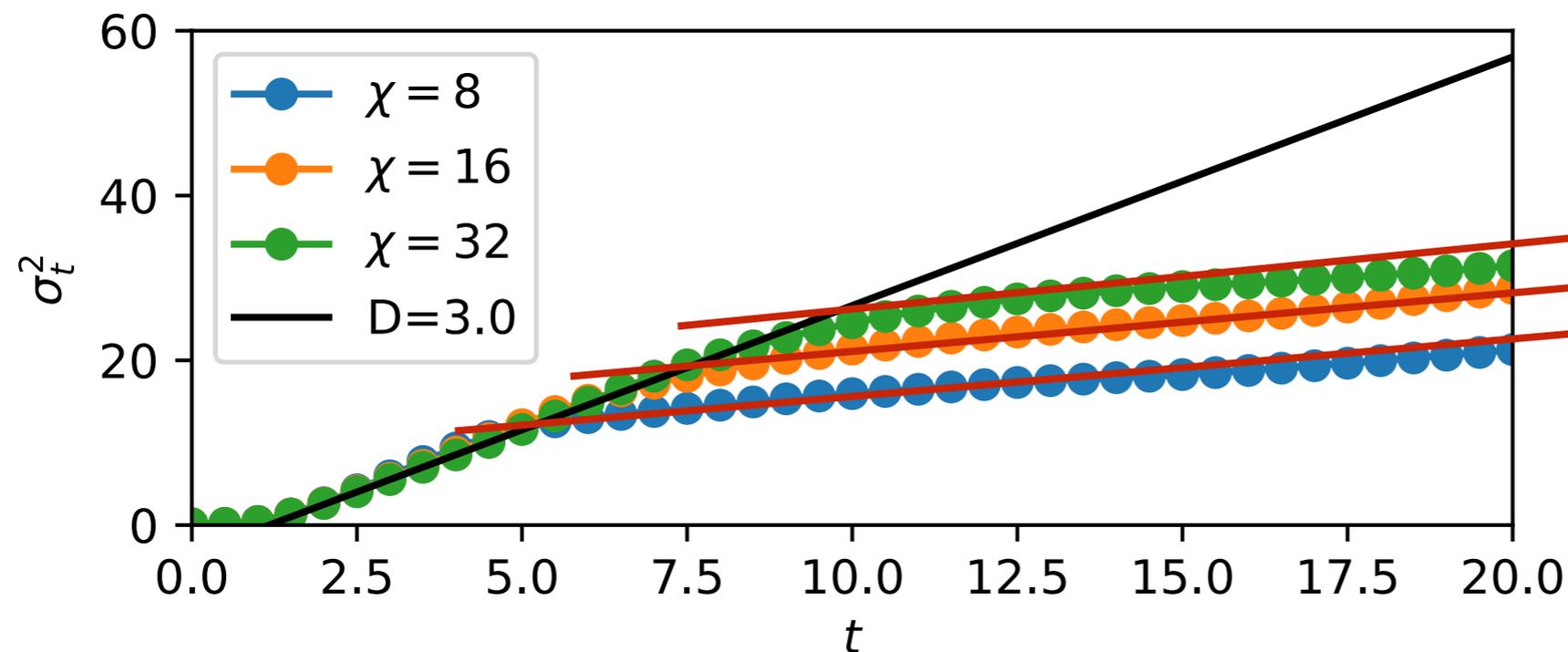


► Global conservation laws (energy, particles,...)

Time-dependent variational principle (TDVP)

XXZ Model with longer range interactions

$$H = \sum_{i>j} a^{i-j} (S_i^x S_j^x + S_i^y S_j^y + \Delta S_i^z S_j^z)$$



- Diffusion constant changes when truncation kicks it!

Dissipation-assisted operator evolution method

Artificial dissipation leads to a decay of operator entanglement, allowing us to capture the dynamics to long times

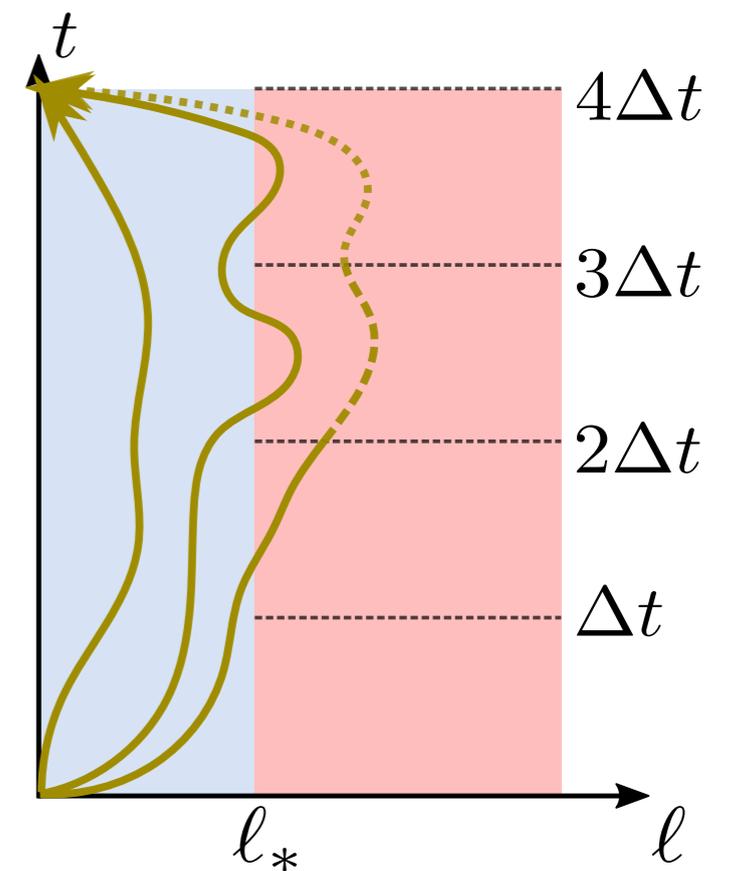
- ▶ Discard information corresponding to n -point functions with $n > \ell_*$.

[Rakovszky, von Keyserlingk, FP, PRB **105**, 07513 (2022)]

- ▶ Errors induced by truncation (“backflow”) are exponentially suppressed in ℓ_*

$$|D - D_{\text{DAOE}}| \sim e^{-\mathcal{O}(\ell_*)}$$

[von Keyserlingk, FP, Rakovszky PRB **105**, 245101 (2022)]



Artificial dissipation that not affects hydrodynamics

$$C(x, t) \equiv \langle q_x(t)q_0(0) \rangle_{\beta=0} = \langle q_x | e^{i\mathcal{L}t} | q_0 \rangle, \quad \mathcal{L} | q_x \rangle \equiv [H, q_x] = -i\partial_t | q_x \rangle$$

Basis of operators: Pauli strings ($\mathbb{1}, X, Y, Z$)

$$\mathcal{S} = \dots ZX\mathbb{1}YX\mathbb{1}\mathbb{1}Y\dots \longrightarrow |q_0(t)\rangle = \sum_{\mathcal{S}} a_{\mathcal{S}} |\mathcal{S}\rangle$$

Artificial Dissipator:

$$\mathcal{D}_{\ell_*, \gamma} |\mathcal{S}\rangle = \begin{cases} |\mathcal{S}\rangle & \text{if } \ell_{\mathcal{S}} \leq \ell_* \\ e^{-\gamma(\ell_{\mathcal{S}} - \ell_*)} |\mathcal{S}\rangle & \text{otherwise} \end{cases}$$

$\mathbb{1}\mathbb{1}\mathbb{1}\mathbb{1}X\mathbb{1}\mathbb{1}$	$\ell_{\mathcal{S}} = 1$
$Y\mathbb{1}\mathbb{1}\mathbb{1}X\mathbb{1}\mathbb{1}$	$\ell_{\mathcal{S}} = 2$
$Y\mathbb{1}Z\mathbb{1}X\mathbb{1}\mathbb{1}$	$\ell_{\mathcal{S}} = 3$

- ▶ Cutoff length $\ell_* = \#\text{non-trivial Paulis}$
- ▶ Should be larger than support of conserved densities!
- ▶ Dissipation strength: γ

Artificial dissipation that not affects hydrodynamics

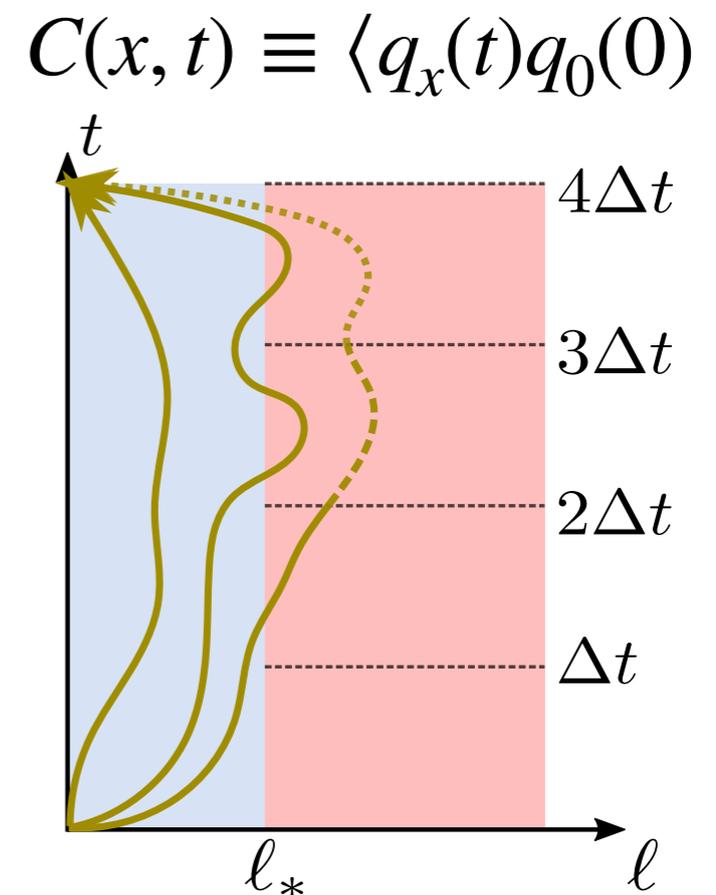
Modified evolution: dissipate after every Δt

$$|\tilde{q}_x(N\Delta t)\rangle \equiv \left(\mathcal{D}_{\ell_*, \gamma} e^{i\mathcal{L}\Delta t} \right)^N |q_x\rangle$$

$$\mathcal{D}_{\ell_*, \gamma} |\mathcal{S}\rangle = \begin{cases} |\mathcal{S}\rangle & \text{if } \ell_{\mathcal{S}} \leq \ell_* \\ e^{-\gamma(\ell_{\mathcal{S}} - \ell_*)} |\mathcal{S}\rangle & \text{otherwise} \end{cases}$$

► Key assumption: **backflow** from long to short operators is weak

(Cf.: **Short memory time** in Zwanzig-Mori memory matrix)

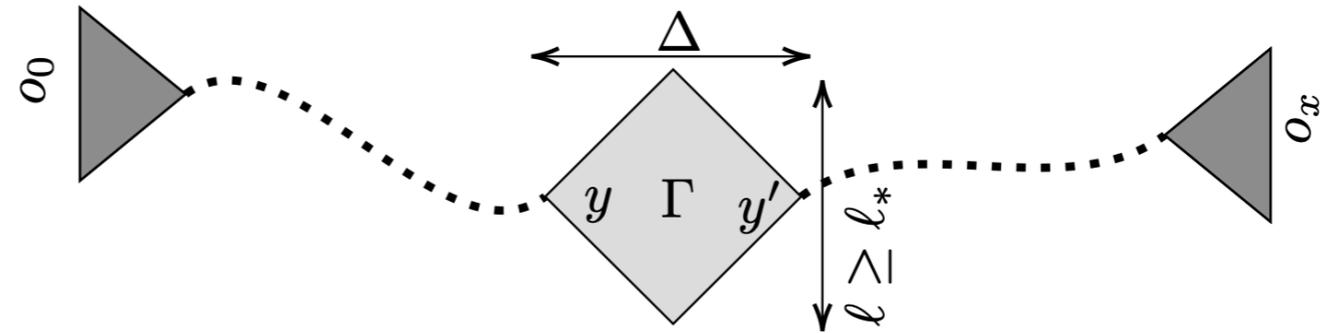


**Reduction of operator entanglement:
Efficient MPS representation!**

Non rigorous theory of “backflow” corrections

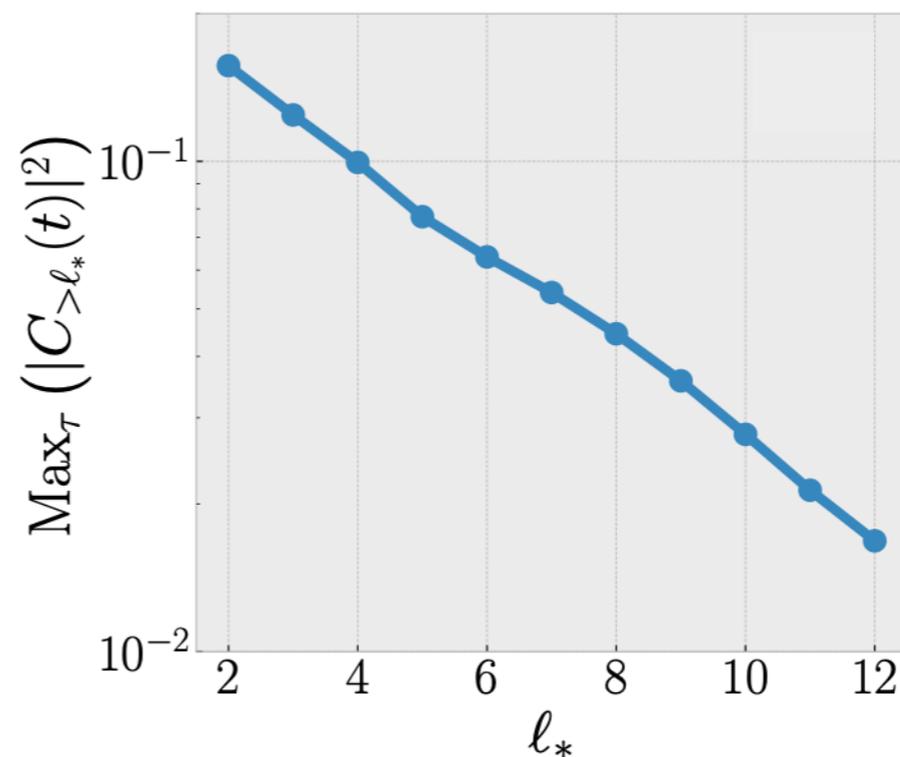
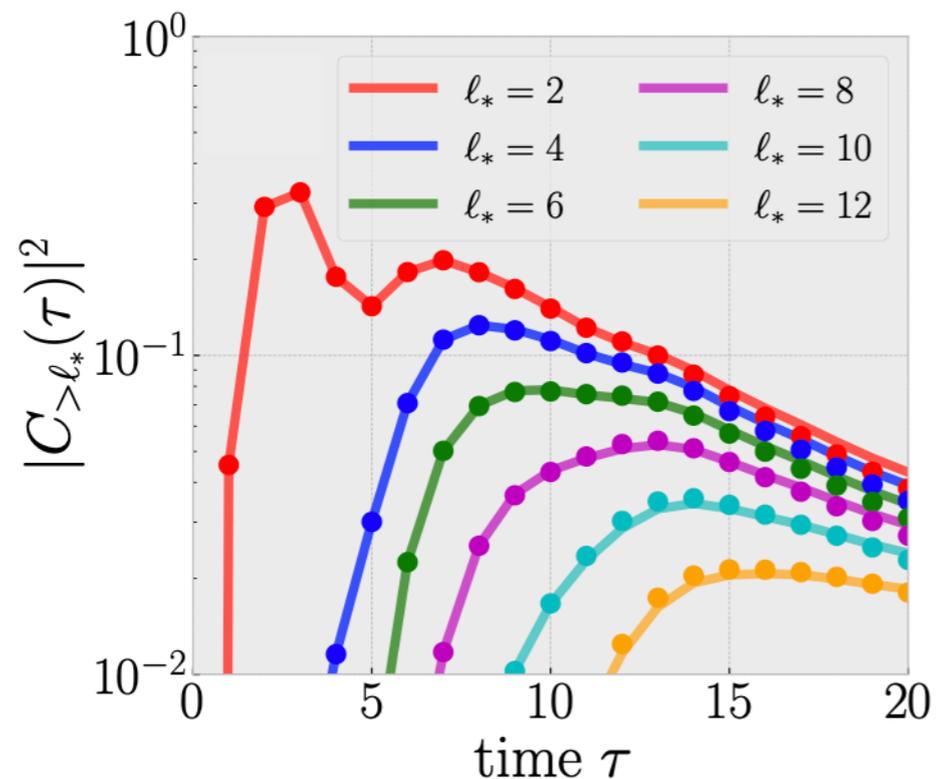
Dynamical correlations

$$C(\tau, x) \equiv \langle o_0(\tau) | o_x \rangle$$



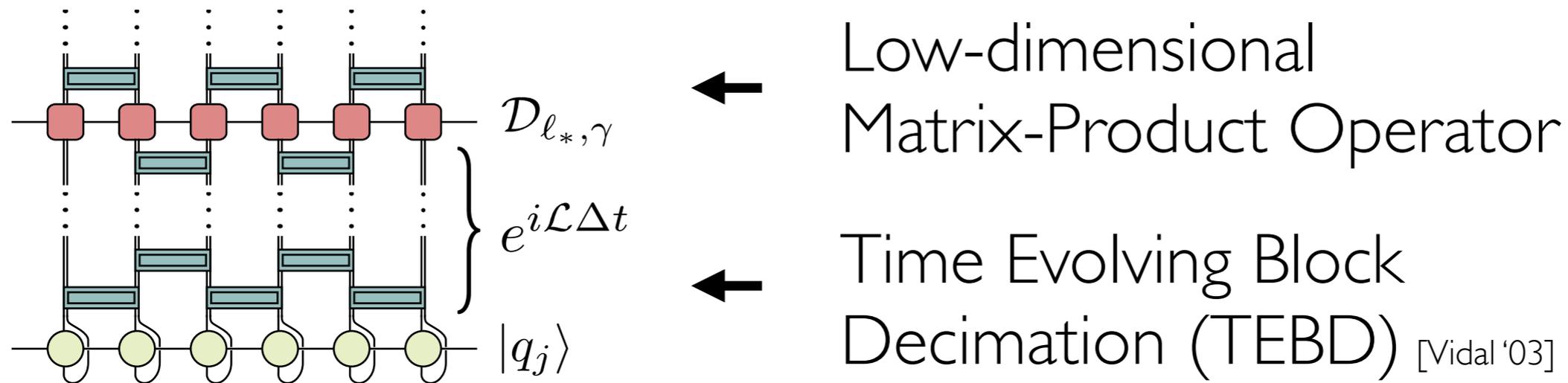
Contributions from “backflow” exponentially

small in l_* : $C_{>l_*}(\tau, x) \equiv \langle o_0 | \mathcal{U}(\tau = 2t, t) \mathcal{P}_{>l_*} \mathcal{U}(t, 0) | o_x \rangle$

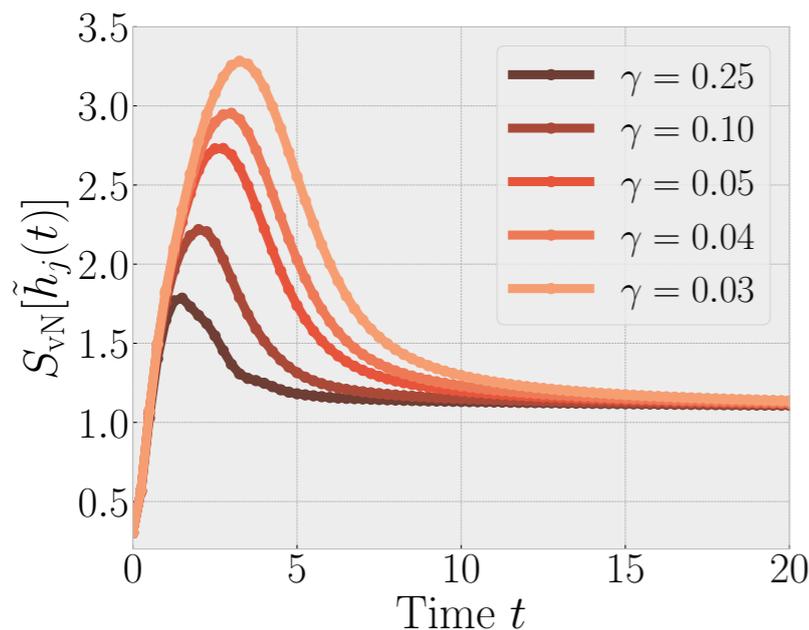


Dissipation stops growth of operator entanglement

Represent dissipative evolution as tensor network



Test on quantum Ising chain:

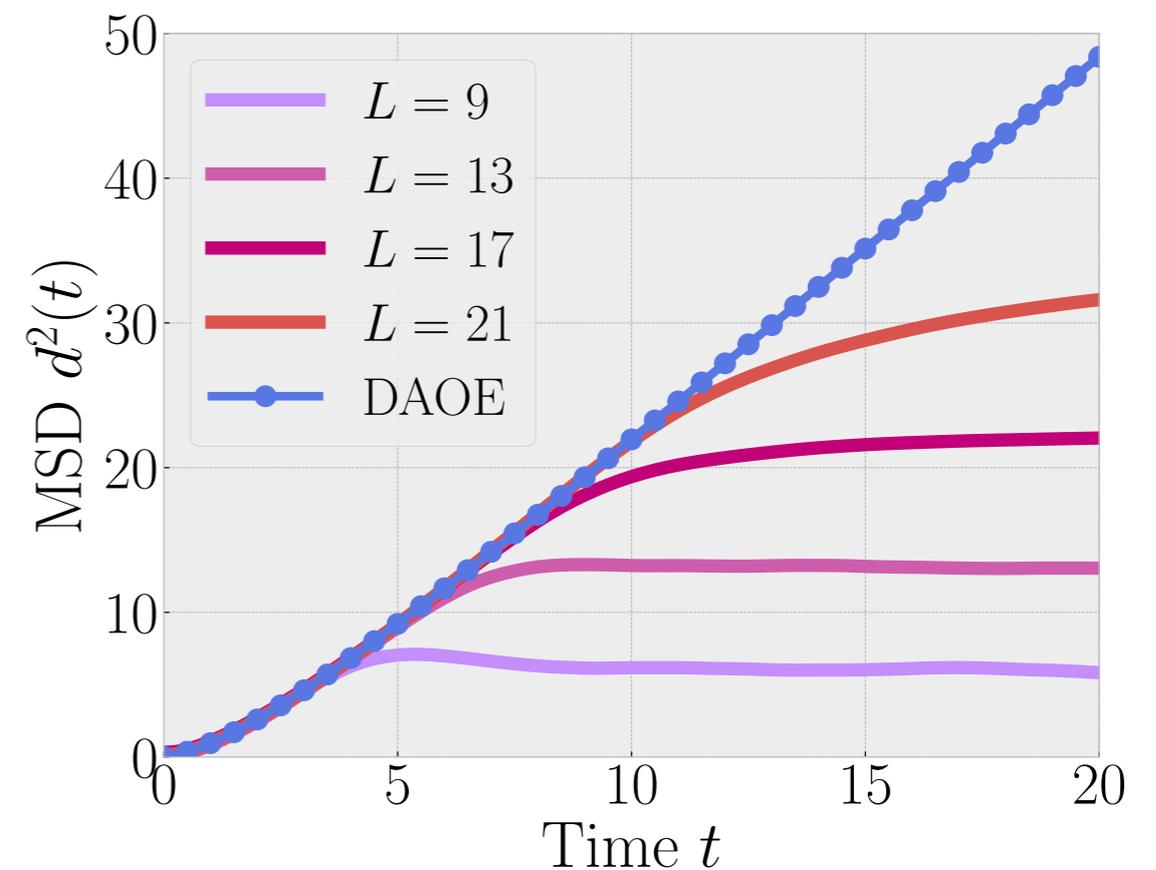
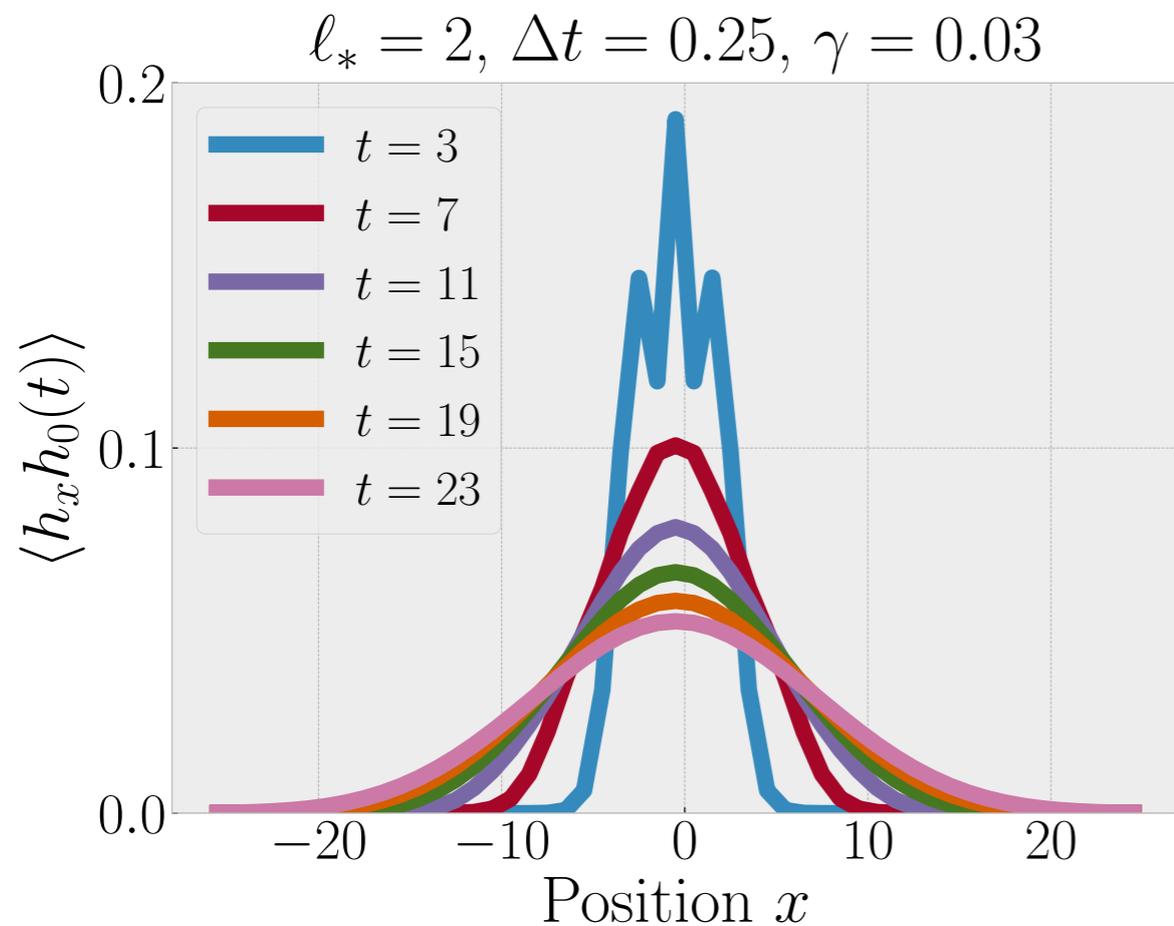


$$H = \sum_j h_j \equiv \sum_j g_x X_j + g_z Z_j + (Z_{j-1} Z_j + Z_j Z_{j+1})/2$$

$$g_x = 1.4; \quad g_z = 0.9045$$

Diffusion constant from mean-square displacement

$$C(x, t) \equiv \langle q_x | \tilde{q}_0(t) \rangle \quad \longrightarrow \quad d^2(t) \equiv \sum_x C(x, t) x^2 \quad (\text{MSD})$$

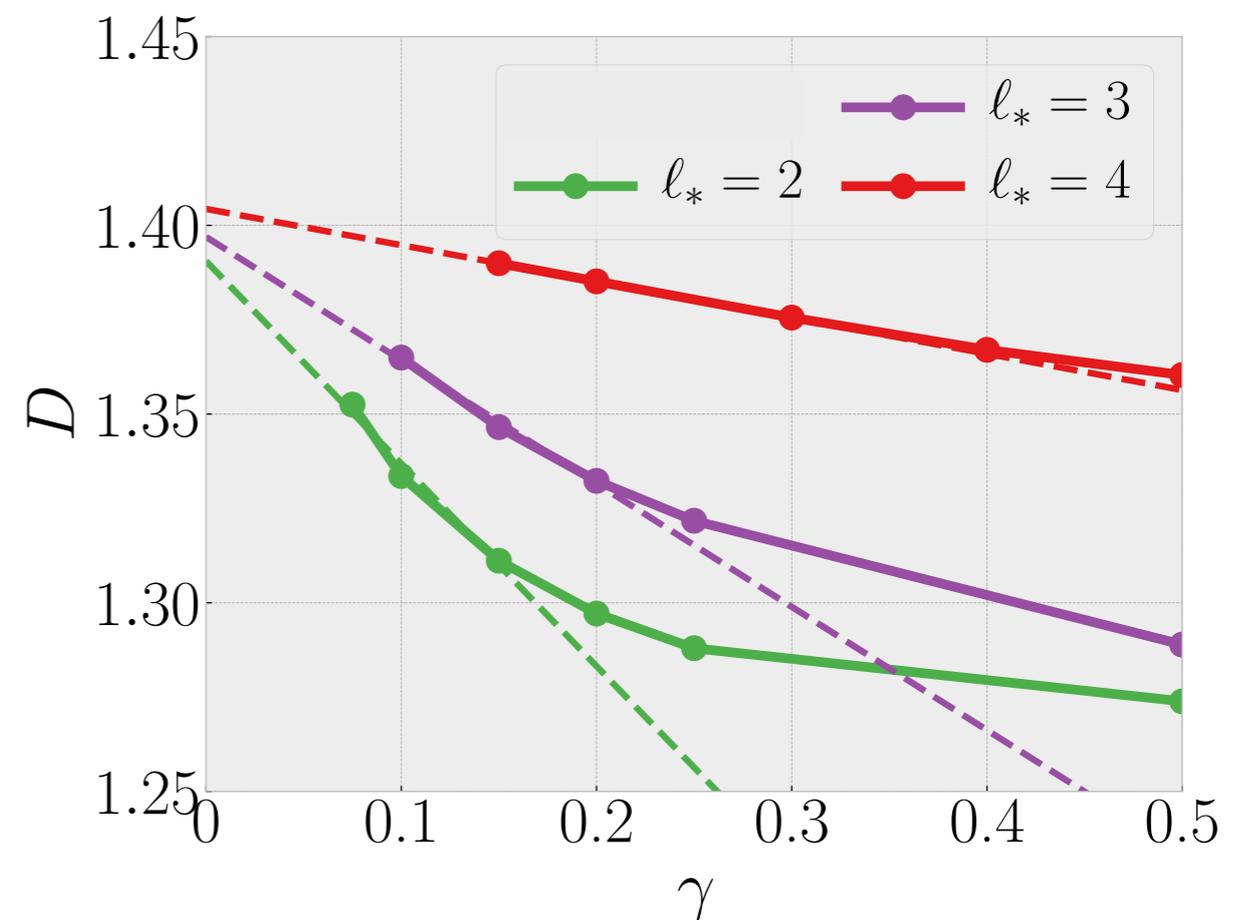
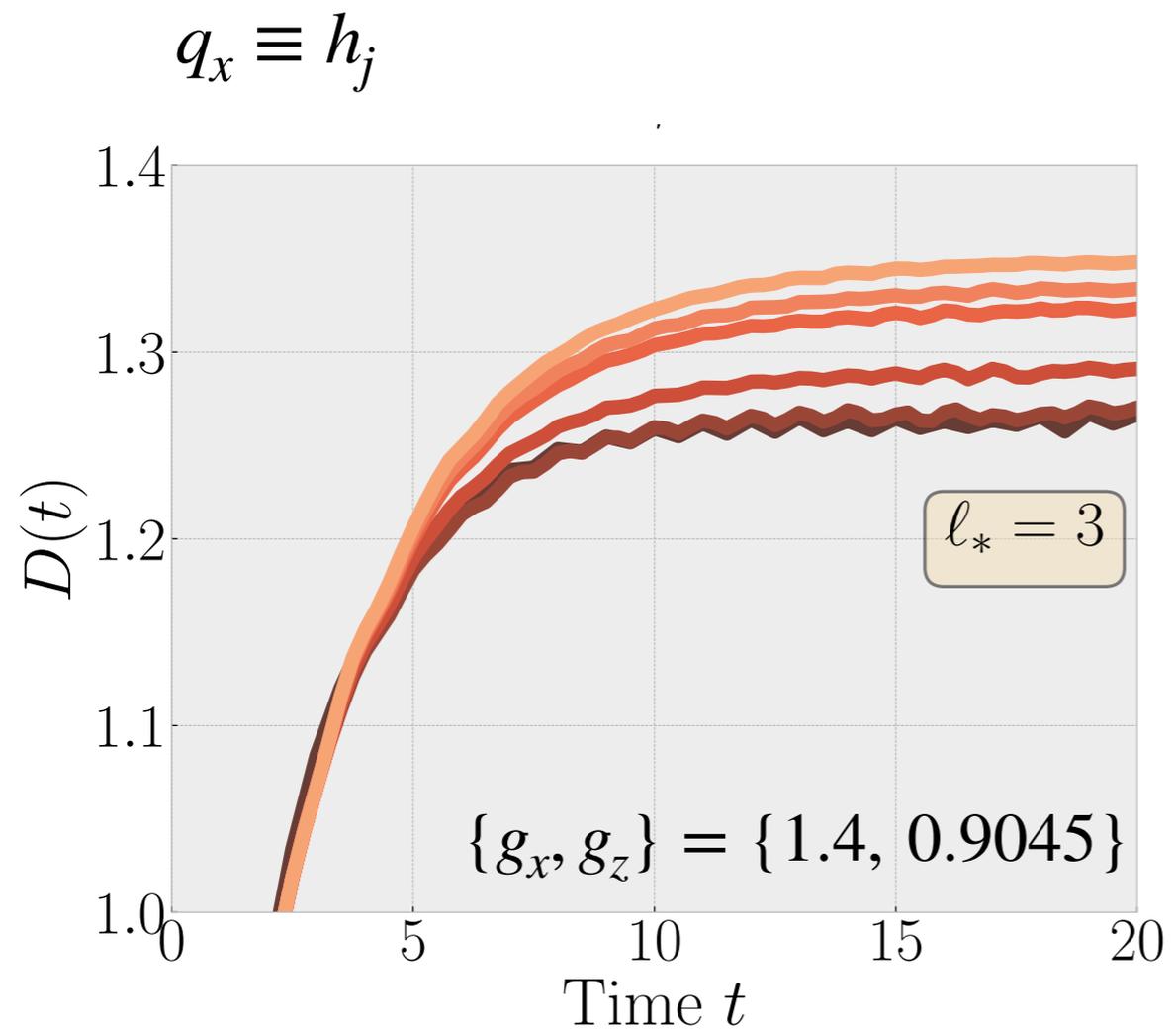


Time-dependent diffusion constant: $2D(t) \equiv \frac{\partial d^2(t)}{\partial t}$

Diffusive transport: $D \equiv \lim_{t \rightarrow \infty} D(t)$

High precision in various models

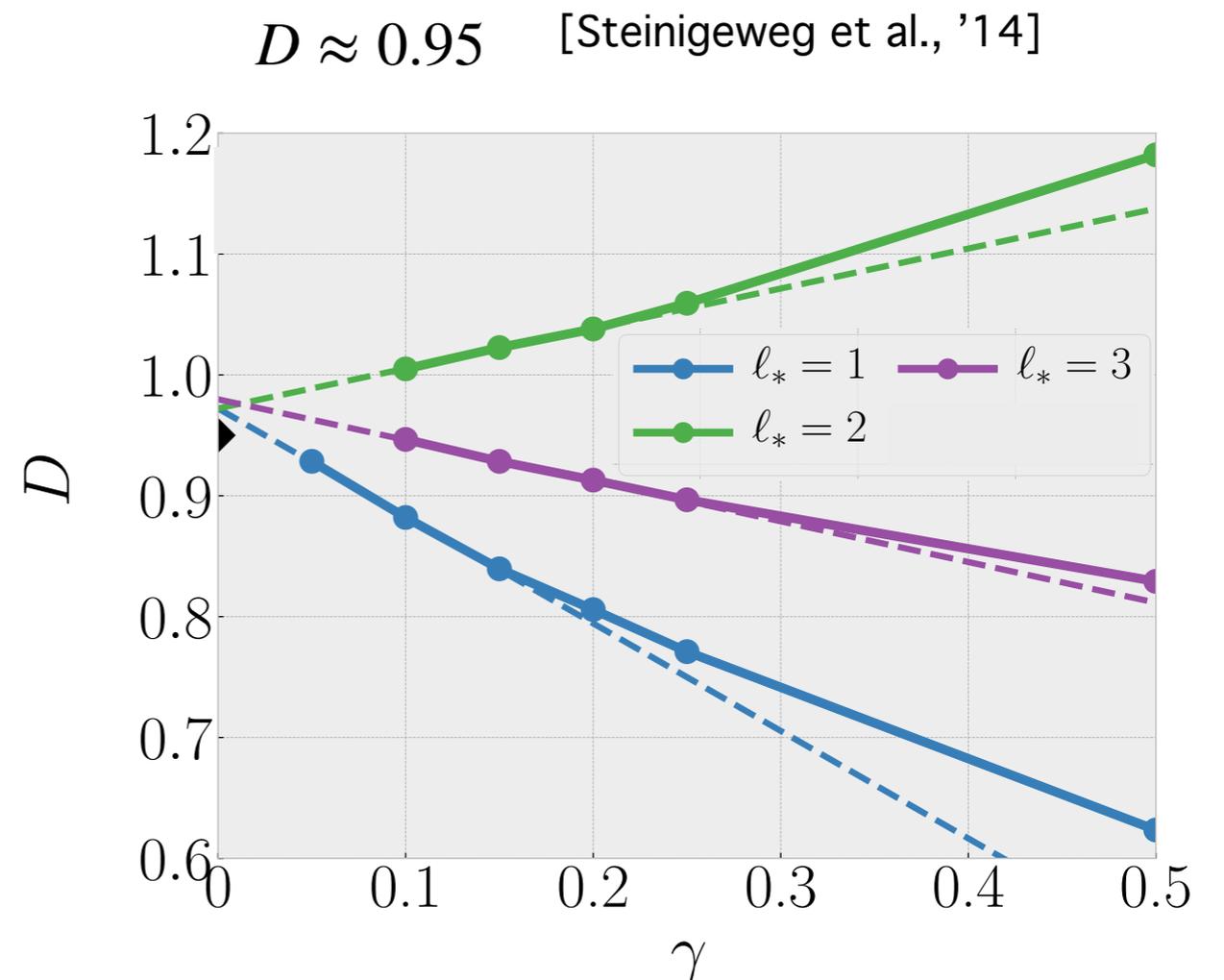
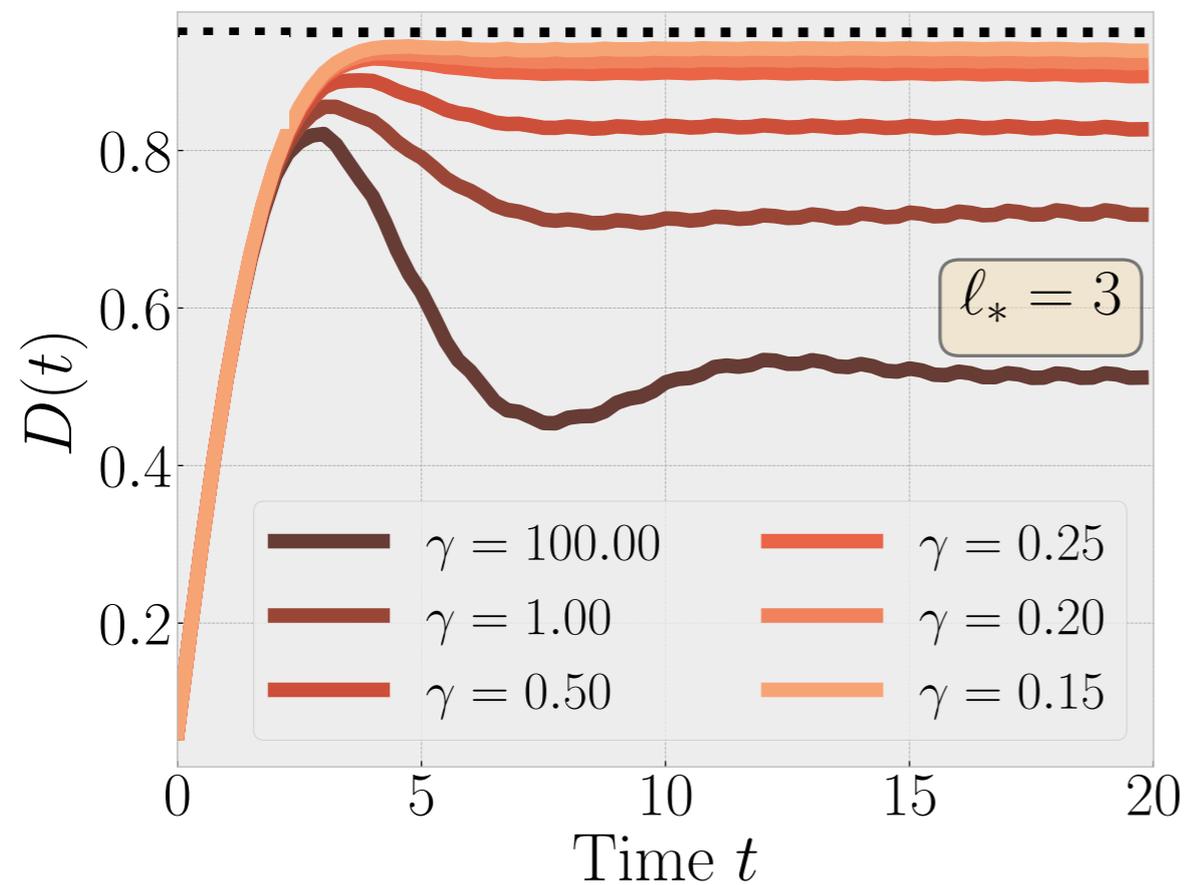
Ising: $H = \sum_j h_j \equiv \sum_j g_x X_j + g_z Z_j + \frac{1}{2}(Z_{j-1}Z_j + Z_jZ_{j+1})$



High precision in various models

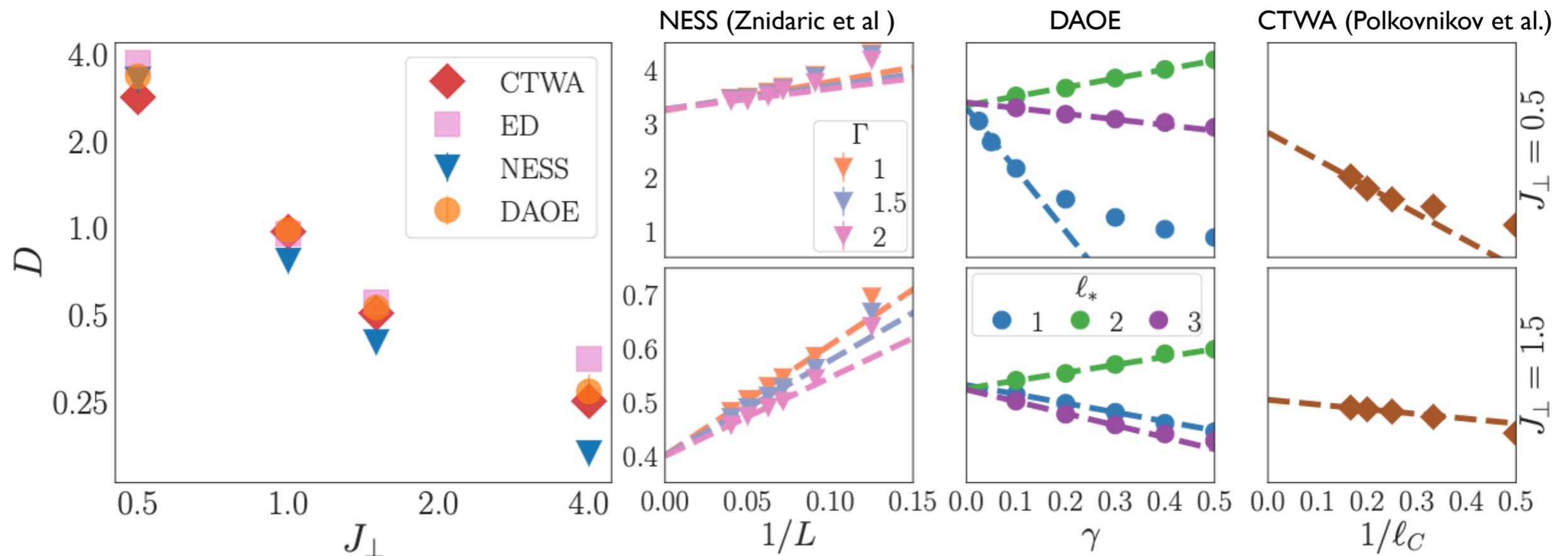
XX ladder:
$$H = \sum_{j=1}^L \sum_{a=1,2} \left(X_{j,a} X_{j+1,a} + Y_{j,a} Y_{j+1,a} \right) + \sum_{j=1}^L \left(X_{j,1} X_{j,2} + Y_{j,1} Y_{j,2} \right)$$

$$q_x \equiv (Z_{j,1} + Z_{j,2})/2$$



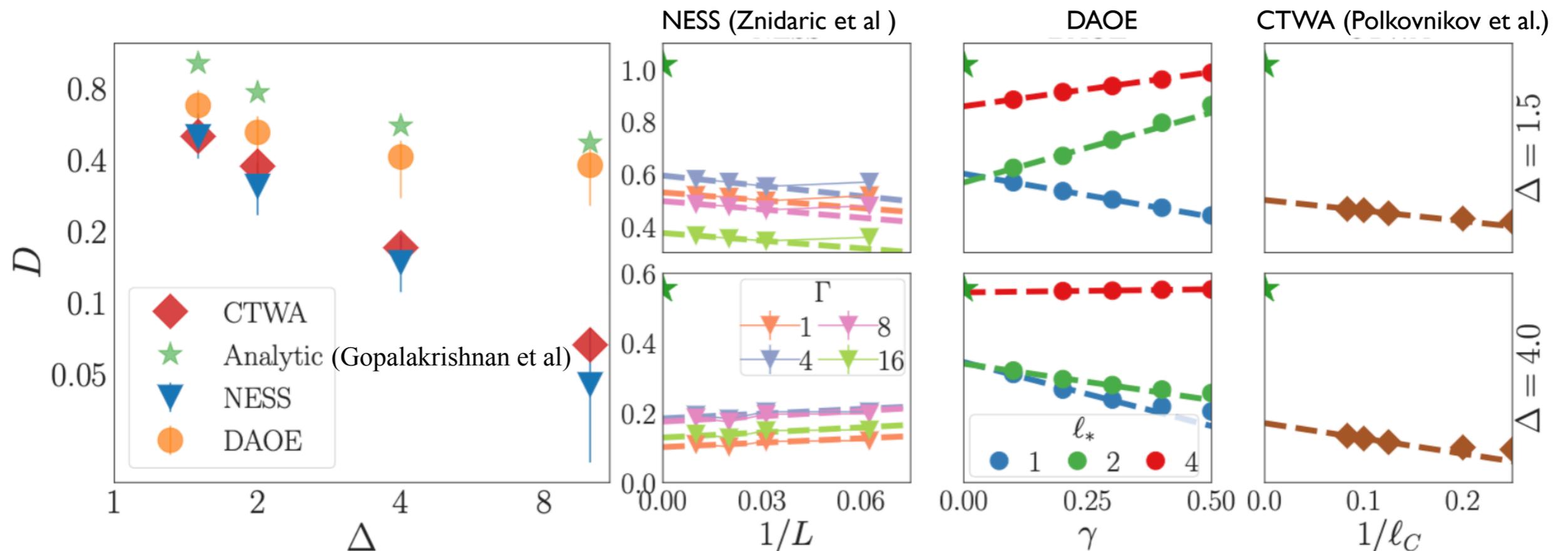
Benchmark to other methods (work in progress)

XX ladder:
$$H = \sum_{j=1}^L \sum_{a=1,2} \left(X_{j,a} X_{j+1,a} + Y_{j,a} Y_{j+1,a} \right) + J_{\perp} \sum_{j=1}^L \left(X_{j,1} X_{j,2} + Y_{j,1} Y_{j,2} \right)$$



Benchmark to other methods (work in progress)

XXZ chain $H = \sum_{j=1}^L \left(X_j X_{j+1} + Y_j Y_{j+1} + \Delta Z_j X_{j+1} \right)$



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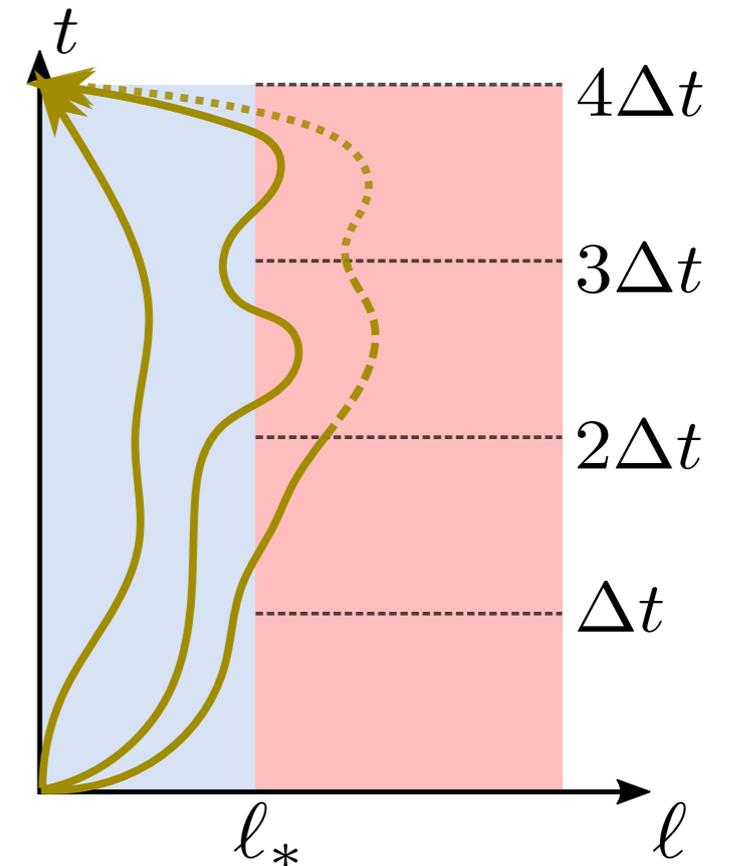
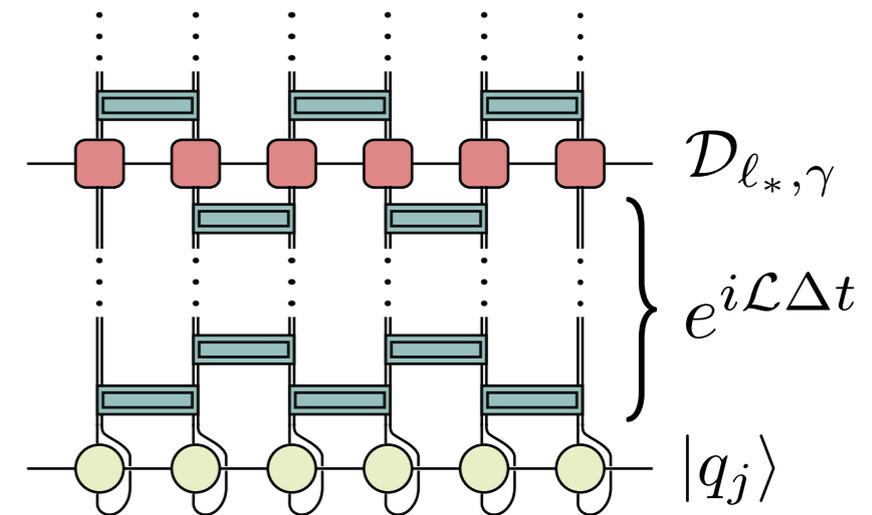
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[von Keyserlingk, FP, Rakovszky PRB **105**, 245101 (2022)]



Thank you!

