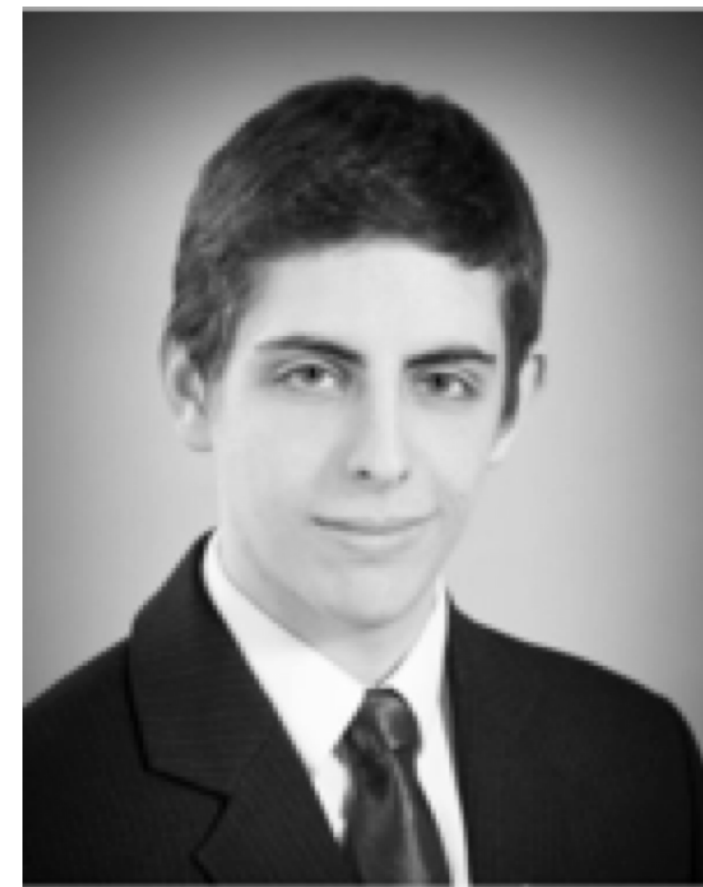


Efficient simulation of quantum transport in 1D

Frank Pollmann

Technische Universität München



T. Rakovszky, Stanford

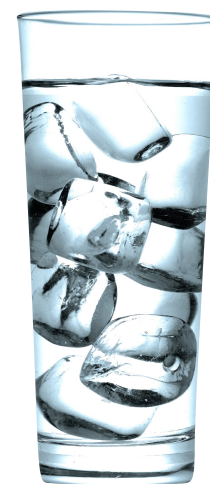


C.v Keyserlingk, Birmingham

[Rakovszky, von Keyserlingk, FP, PRB **105**, 075137 (2022)]

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

Quantum Thermalization



t

$$\rho_{\text{Block}} = \rho_{\text{Thermal}}$$

$$| \text{---} \rangle \quad S = 0$$

$$U_t = \exp(-itH)$$

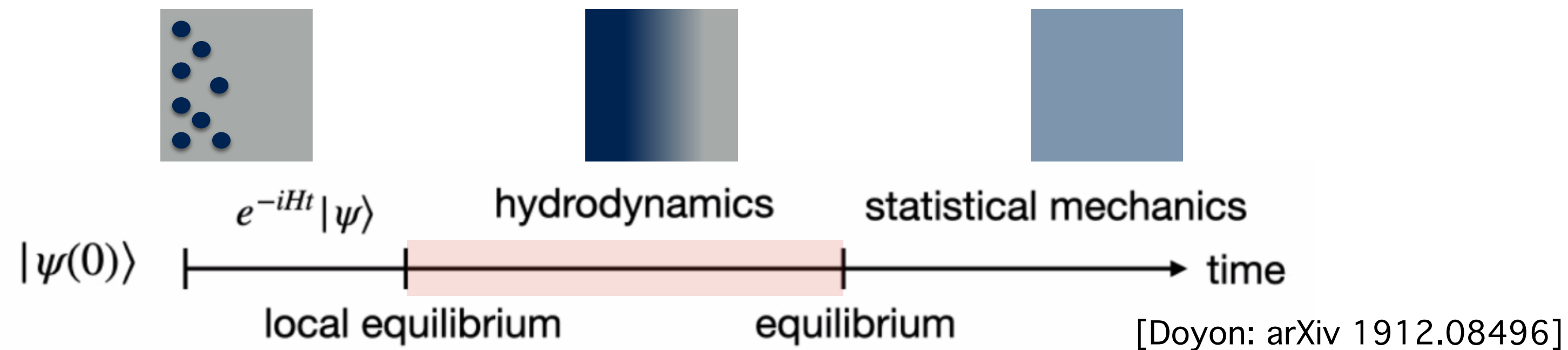
[non-integrable, local Hamiltonian]

$$| \uparrow \downarrow \uparrow \downarrow \uparrow \downarrow \uparrow \downarrow \uparrow \rangle \quad S = 0$$

Closed quantum

Characterizing thermalization dynamics

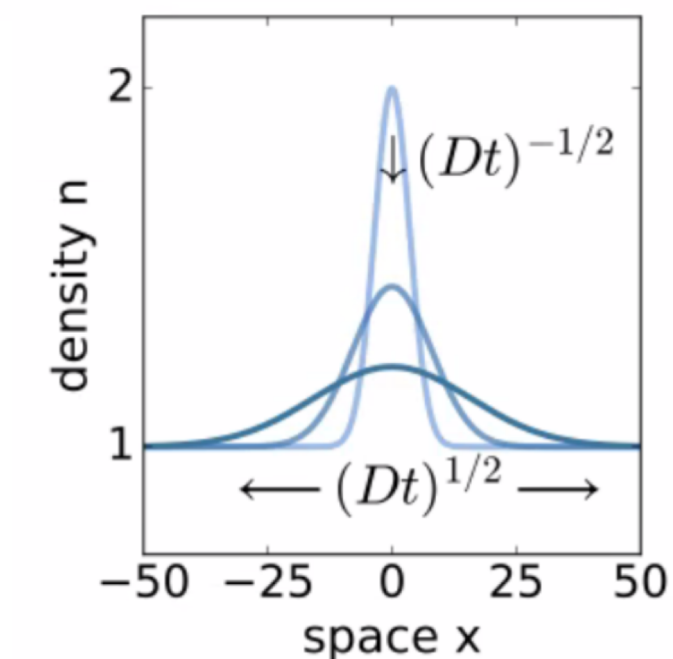
Universal hydrodynamic features tend to emerge in the low-frequency, long-wavelength limit



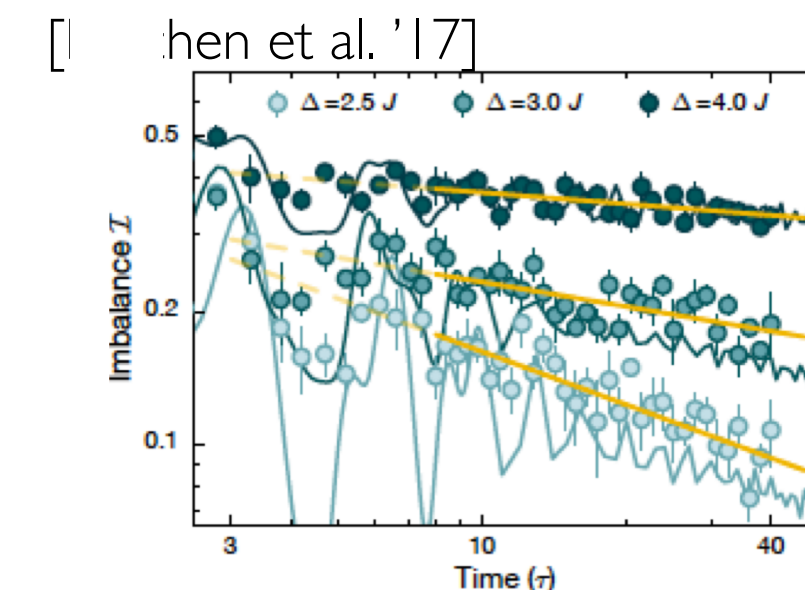
Emergent hydrodynamic relaxation: Diffusion

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

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



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

$$|\psi(t)\rangle = e^{-itH} |\psi(0)\rangle$$

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

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

Matrix-Product States

Low entanglement: Matrix-Product States

[M. Fannes et al. 92]

$$d^L \rightarrow Ld\chi^2$$

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

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

Diagrammatic representation

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

$$A_{\alpha, \beta}^j = \begin{array}{c} j \\ | \\ \bullet \\ | \\ \alpha \quad \beta \end{array} \quad \begin{array}{l} \alpha, \beta = 1 \dots \chi \\ j = 1 \dots d \end{array}$$

Numerical complexity of many-body dynamics

Directly simulate the time evolution within the full many-body Hilbert space

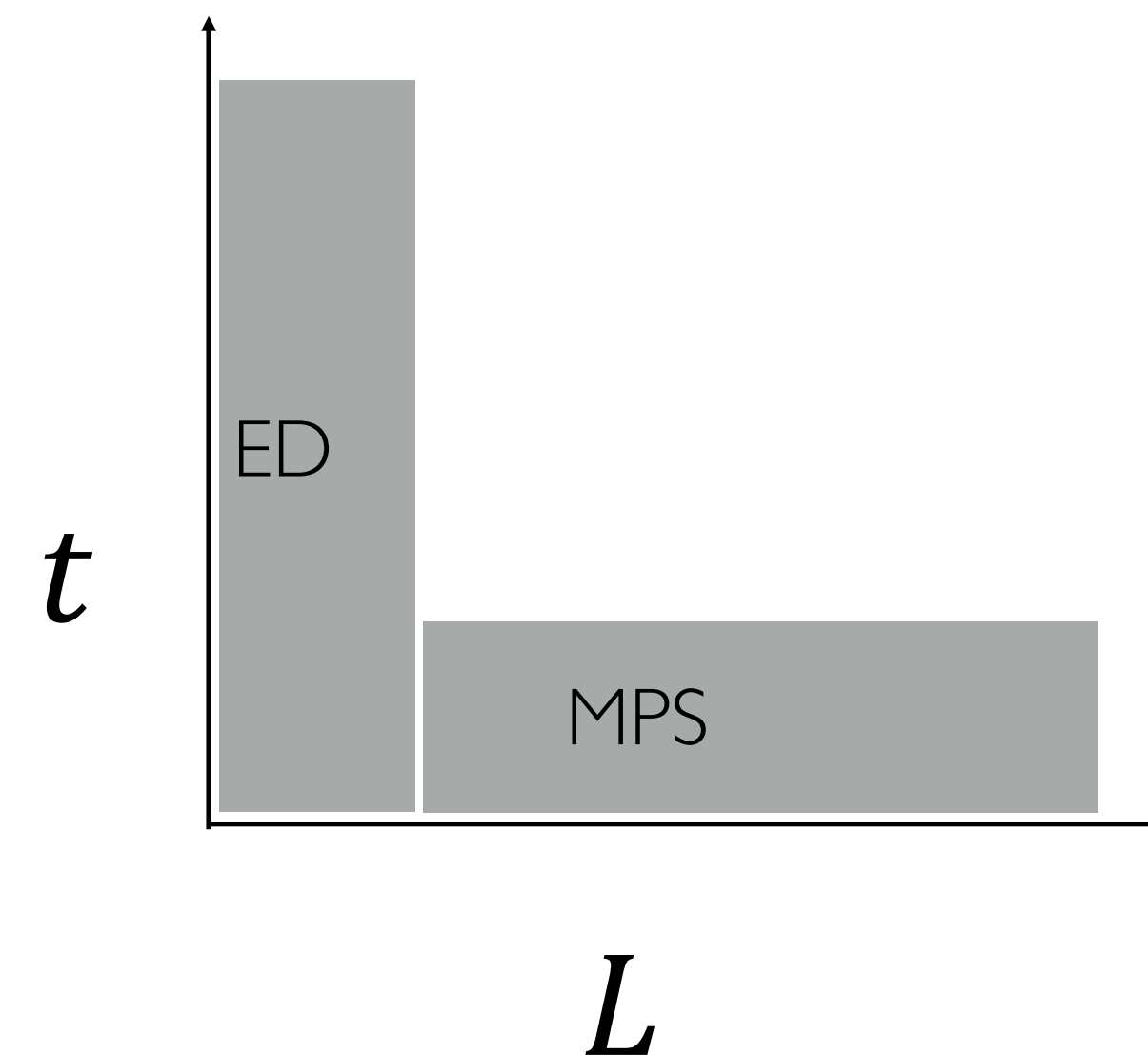
$$|\psi(t)\rangle = e^{-itH}|\psi(0)\rangle$$

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

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

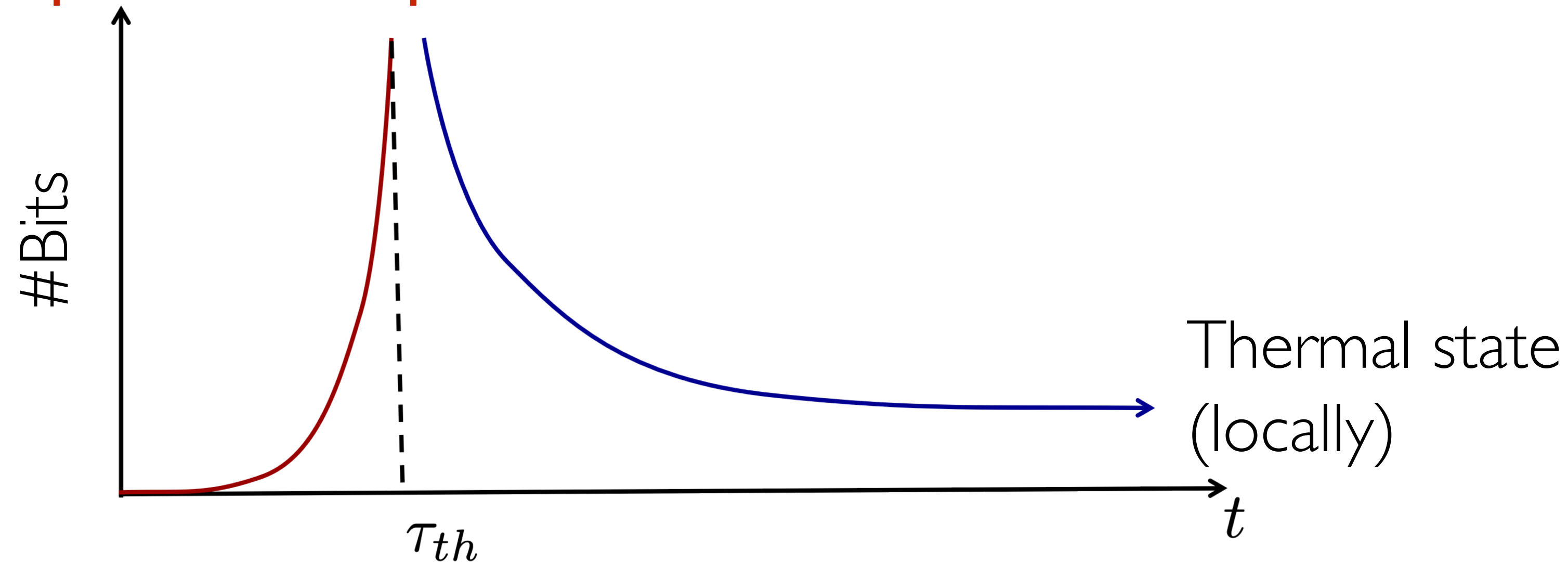
Matrix-Product State based numerics

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



“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]

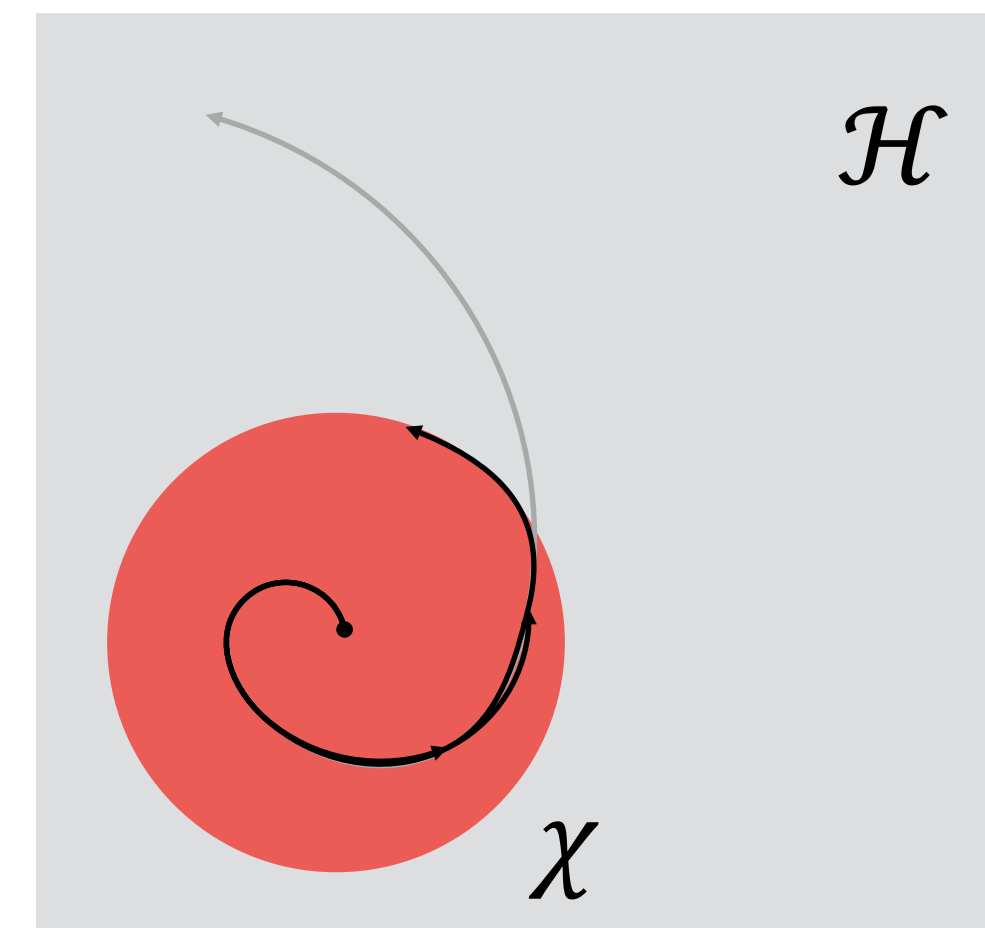
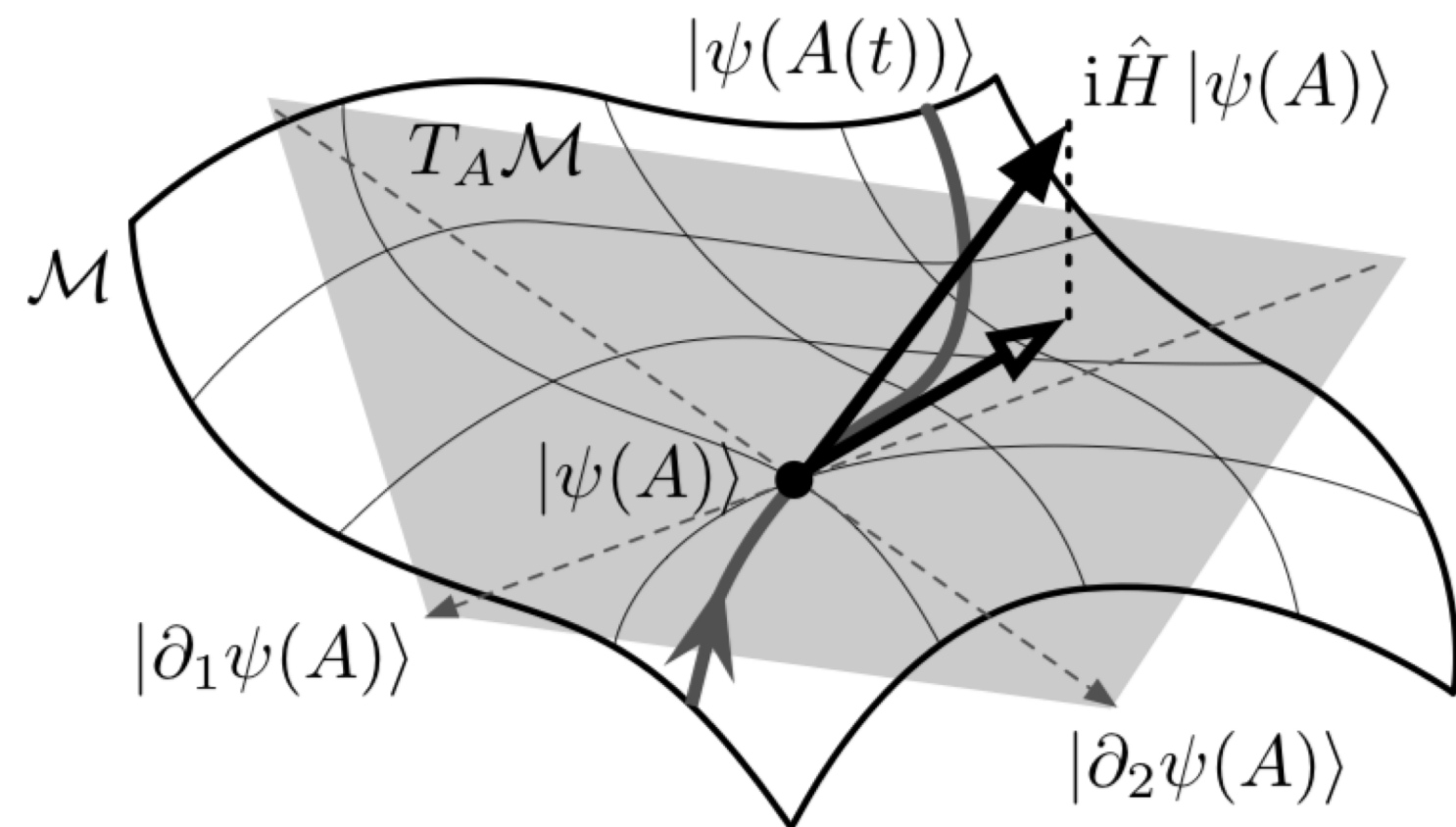
Time-dependent variational principle (TDVP)

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]

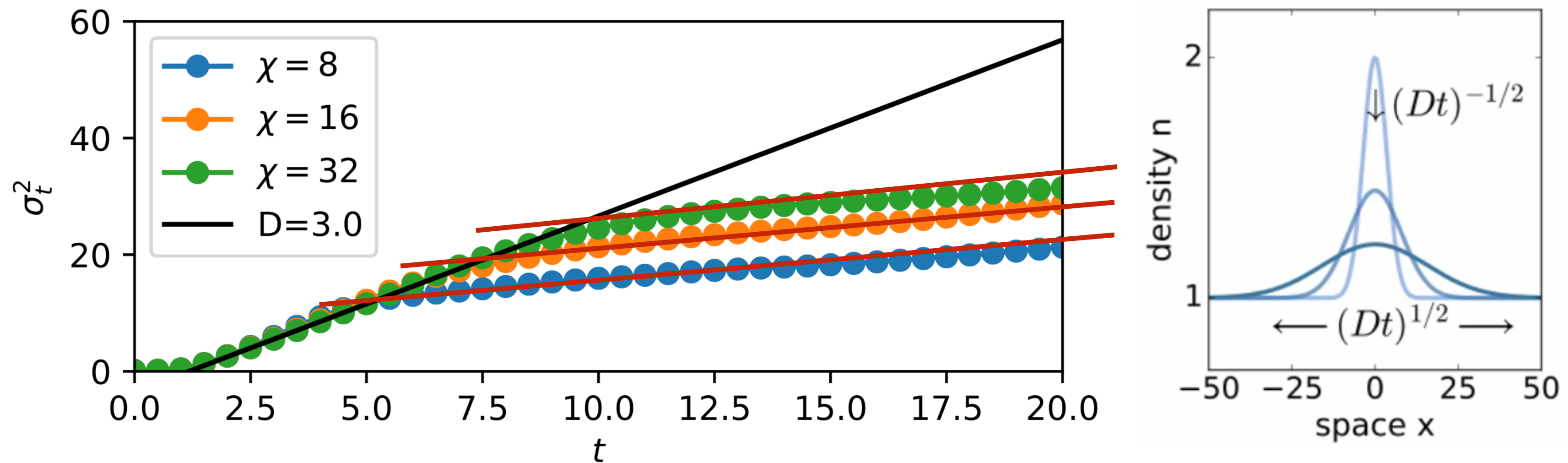


- Global conservation laws (energy, particles,...)

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

Heisenberg Picture: 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 $n > \ell_*$.
- ▶ Errors induced by truncation (“backflow”) are exponentially suppressed in ℓ_*

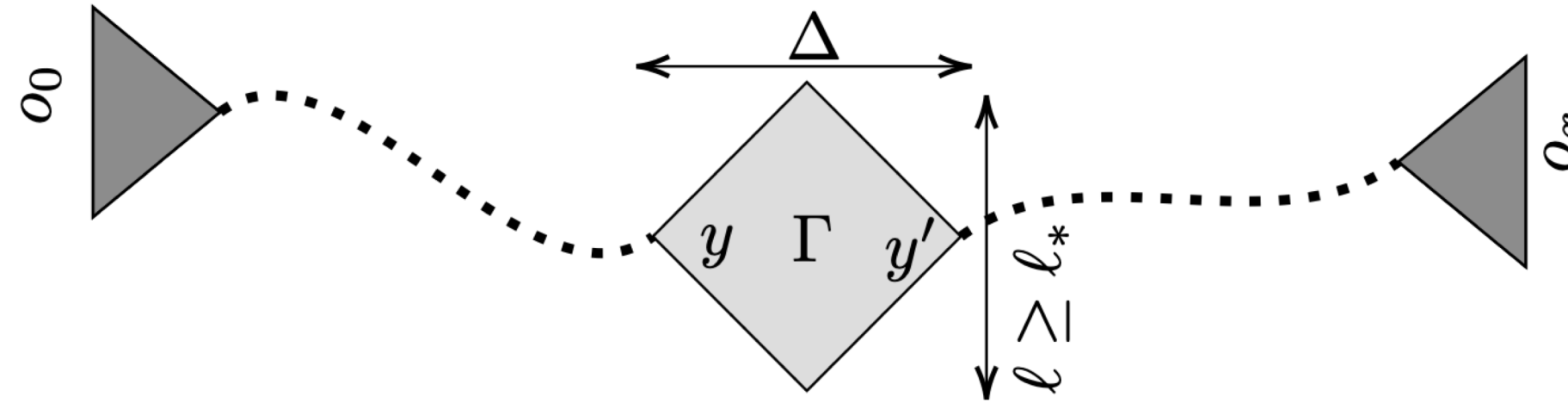
[Rakovszky, von Keyserlingk, FP, PRB **105**, 075137 (2022)]

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

Non rigorous theory of “backflow” corrections

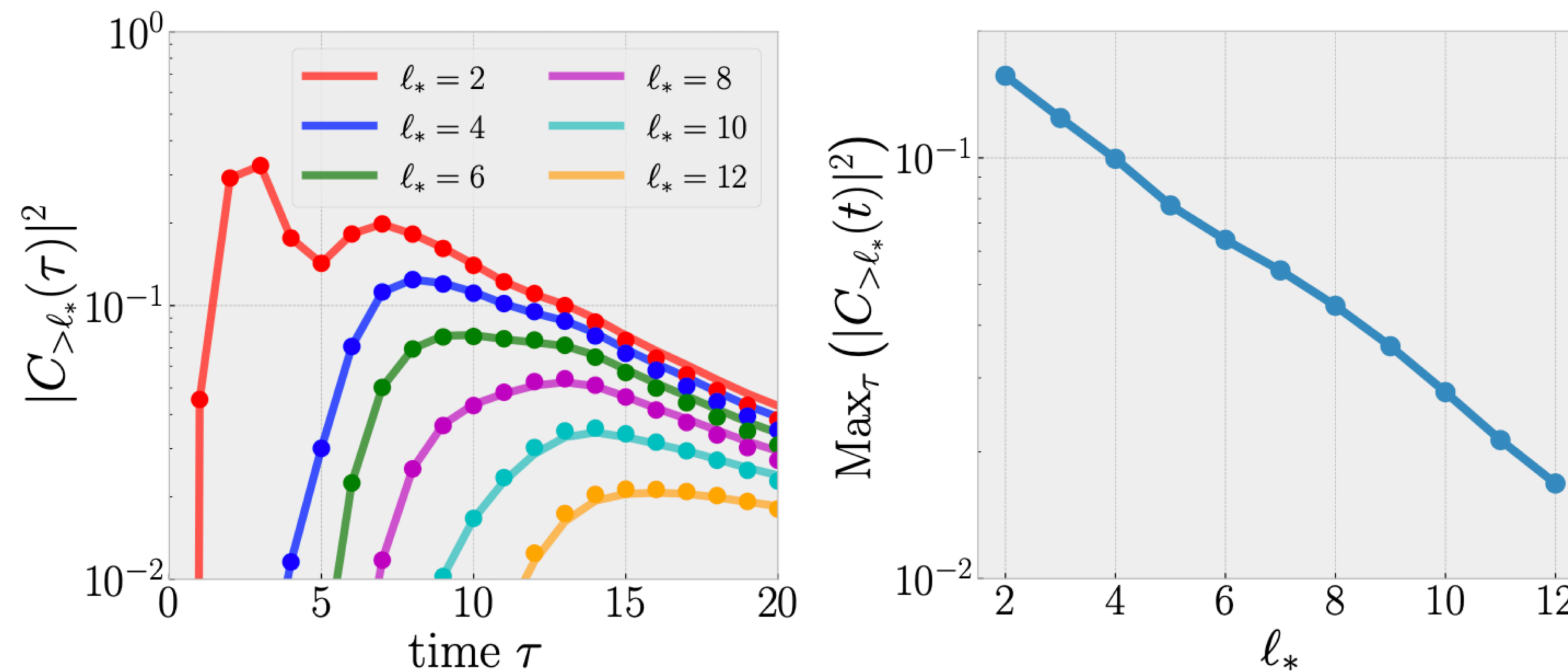
Dynamical correlations

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



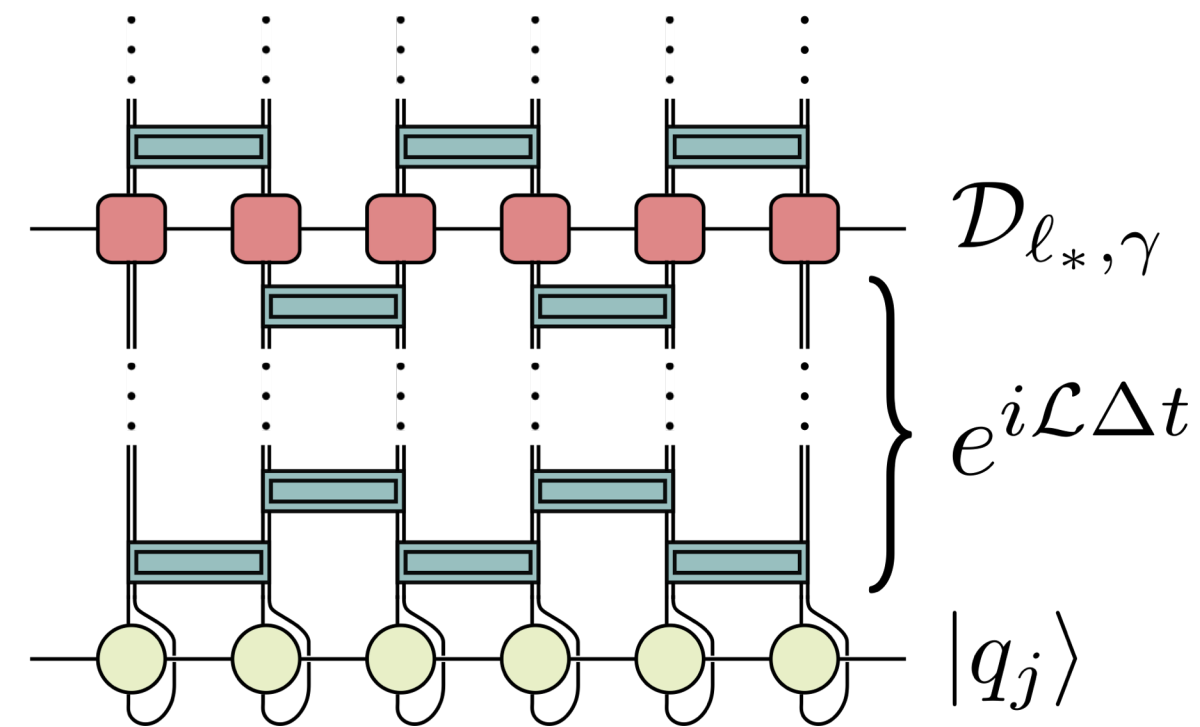
Contributions from “backflow” exponentially

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



Dissipation stops growth of operator entanglement

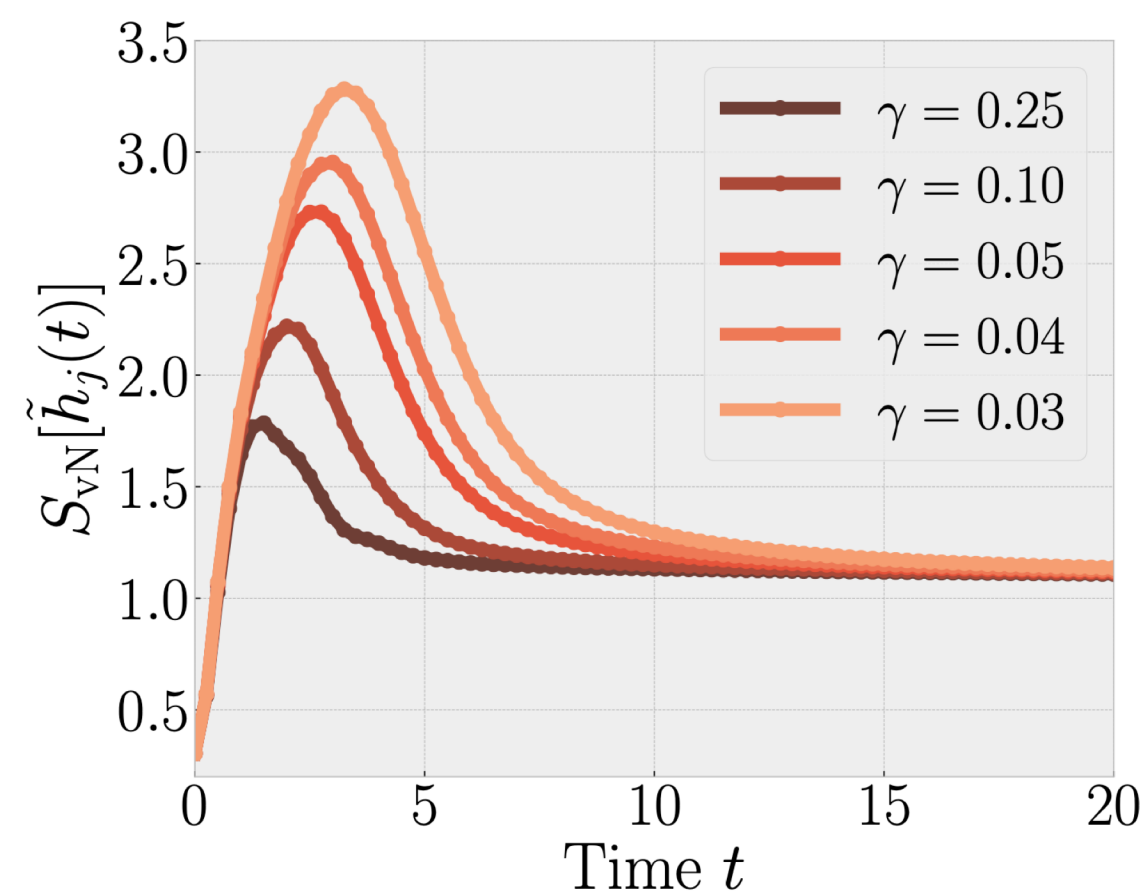
Represent dissipative evolution as tensor network



Low-dimensional
Matrix-Product Operator

Time Evolving Block
Decimation (TEBD) [Vidal '03]

Test on quantum Ising chain:

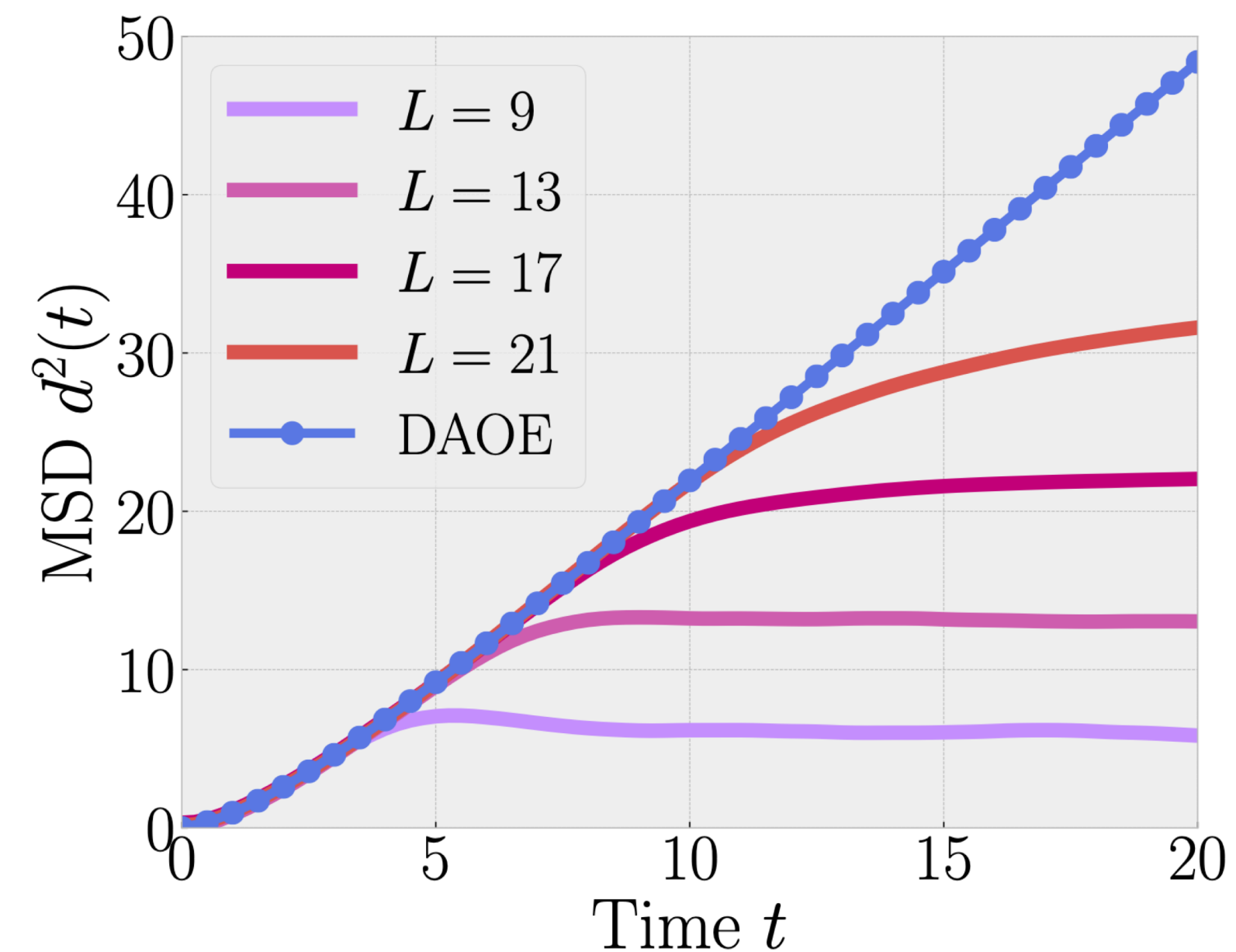
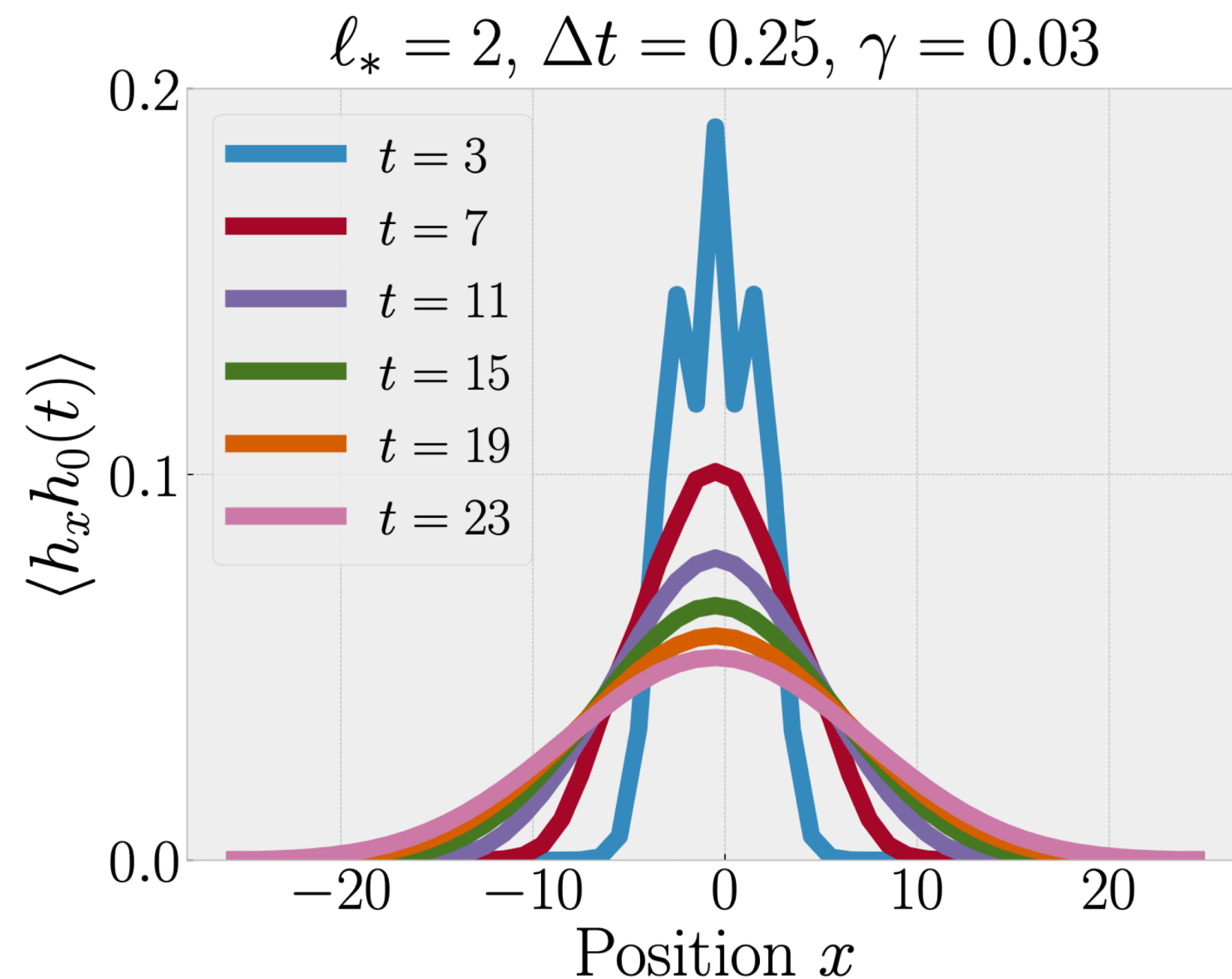


$$H = \sum_j h_j \equiv \sum_j g_x X_j + g_z Z_j + (Z_{j-1} Z_j + Z_j Z_{j+1})/2$$

$$g_x = 1.4; g_z = 0.9045$$

Diffusion constant from mean-square displacement

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



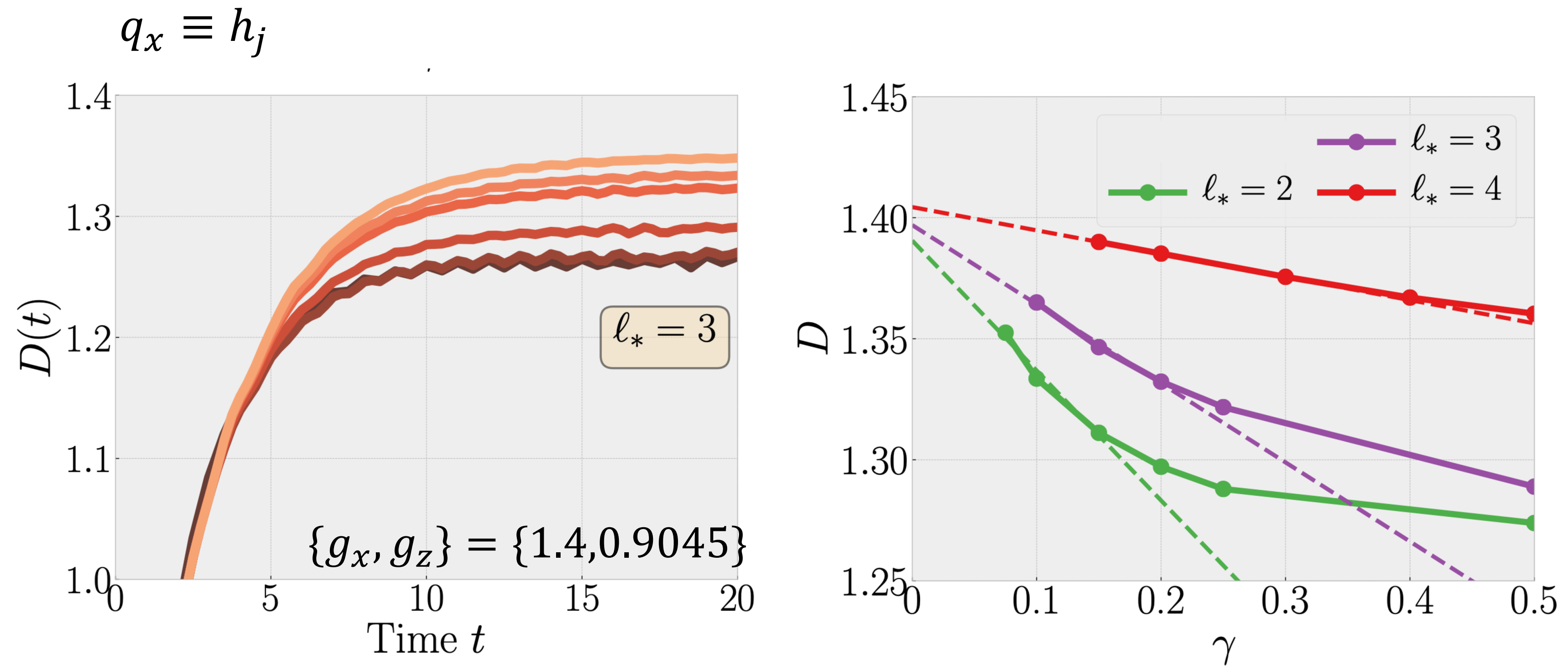
Time-dependent diffusion constant: $2D(t) \equiv \frac{\partial d^2(t)}{\partial t}$

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

[Rakovszky, von Keyserlingk, FP, PRB **105**, 075113 (2022)]

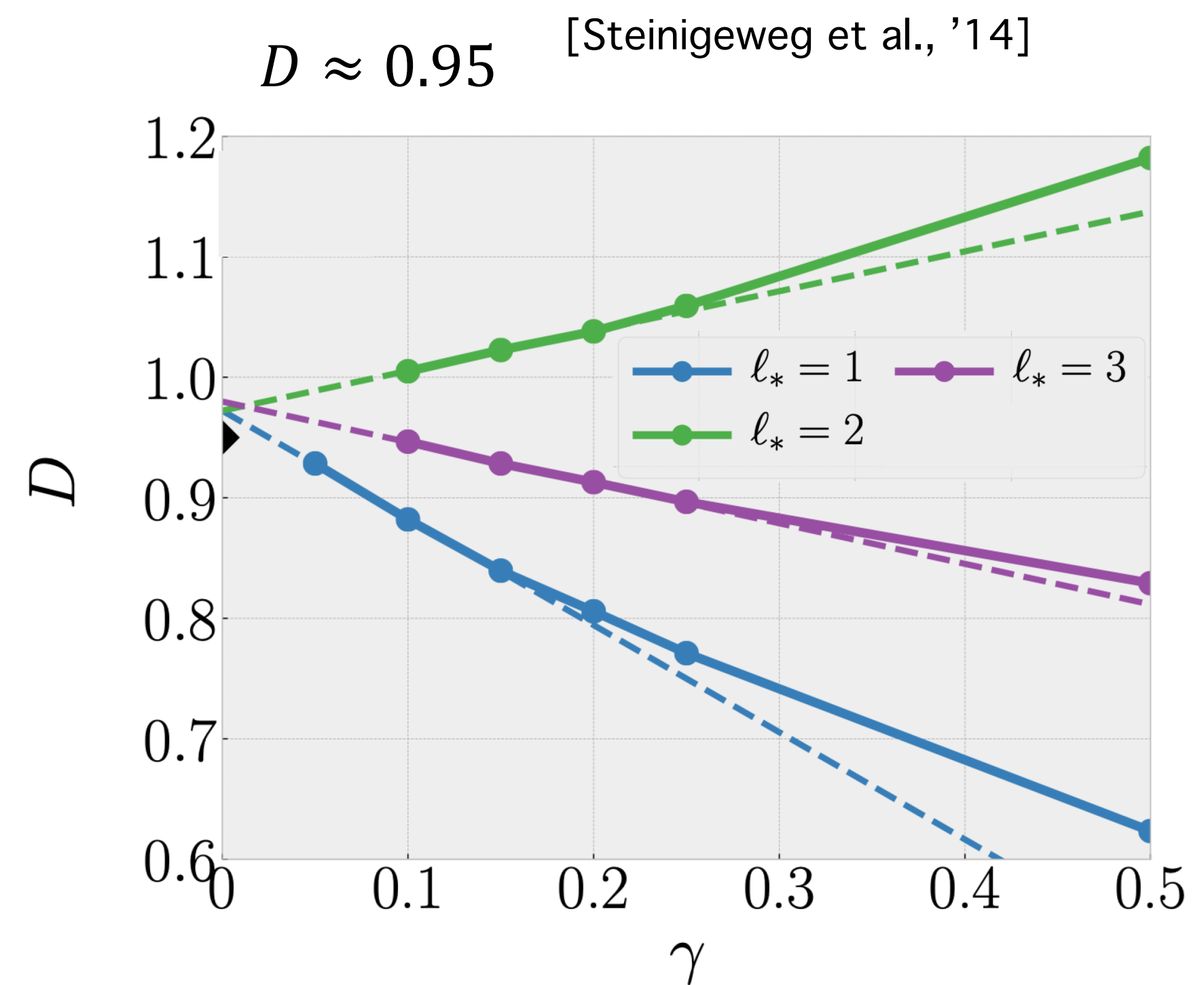
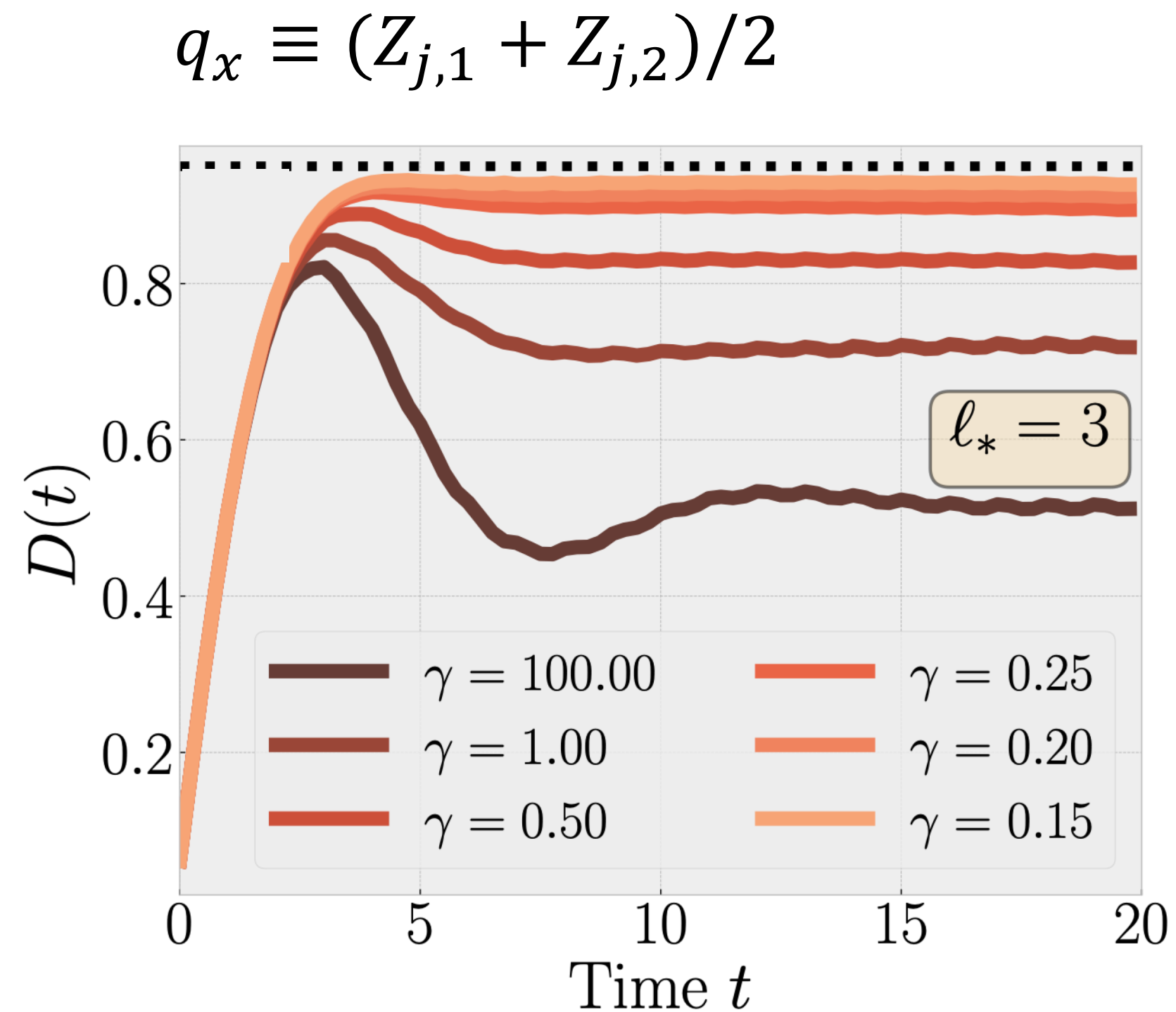
High precision in various models

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



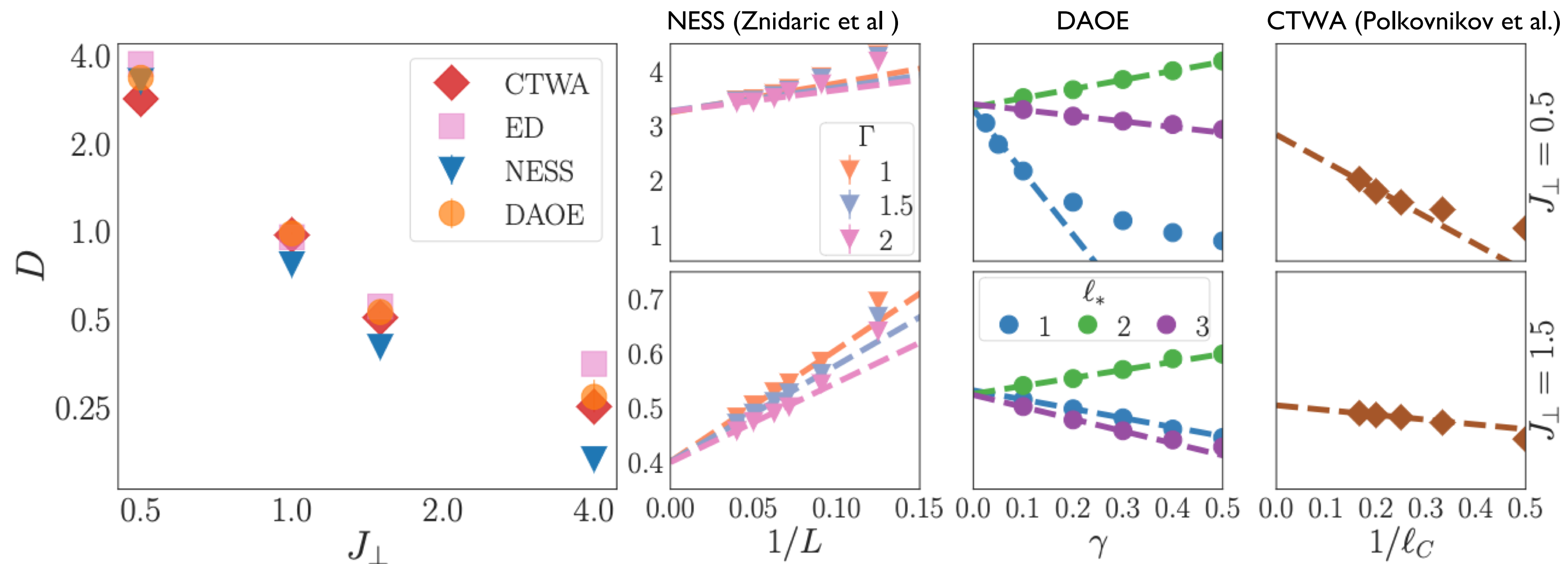
High precision in various models

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



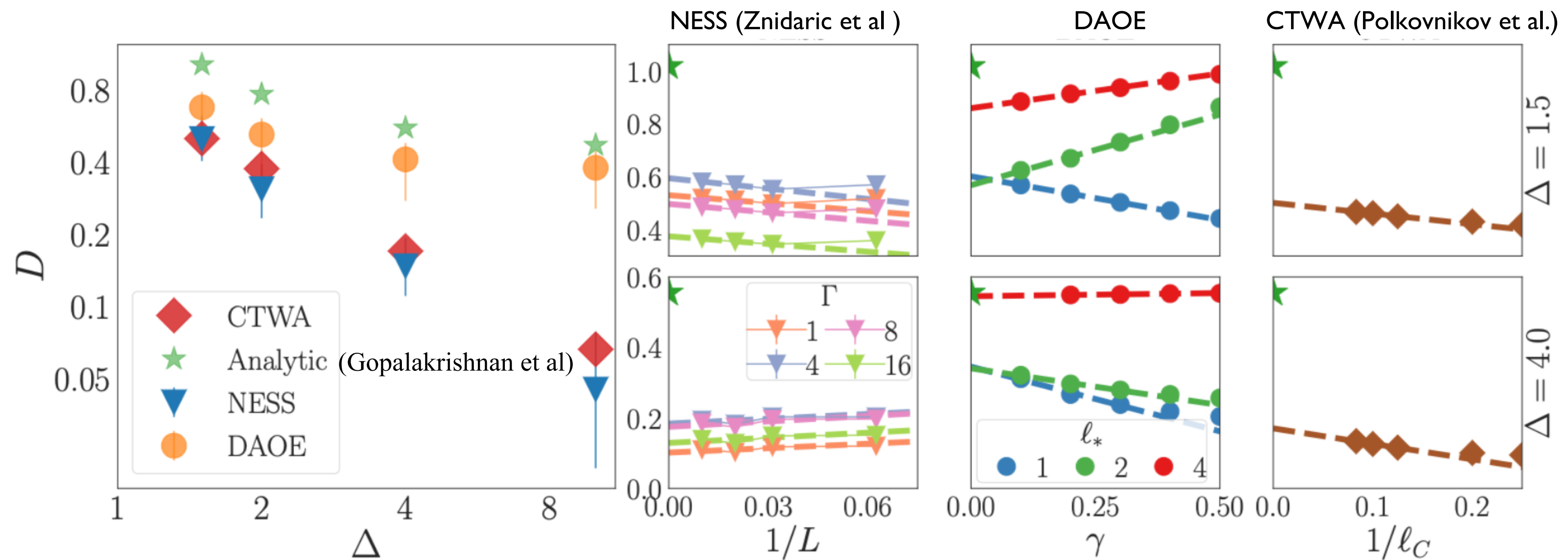
Benchmark to other methods (work in progress)

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



Benchmark to other methods (work in progress)

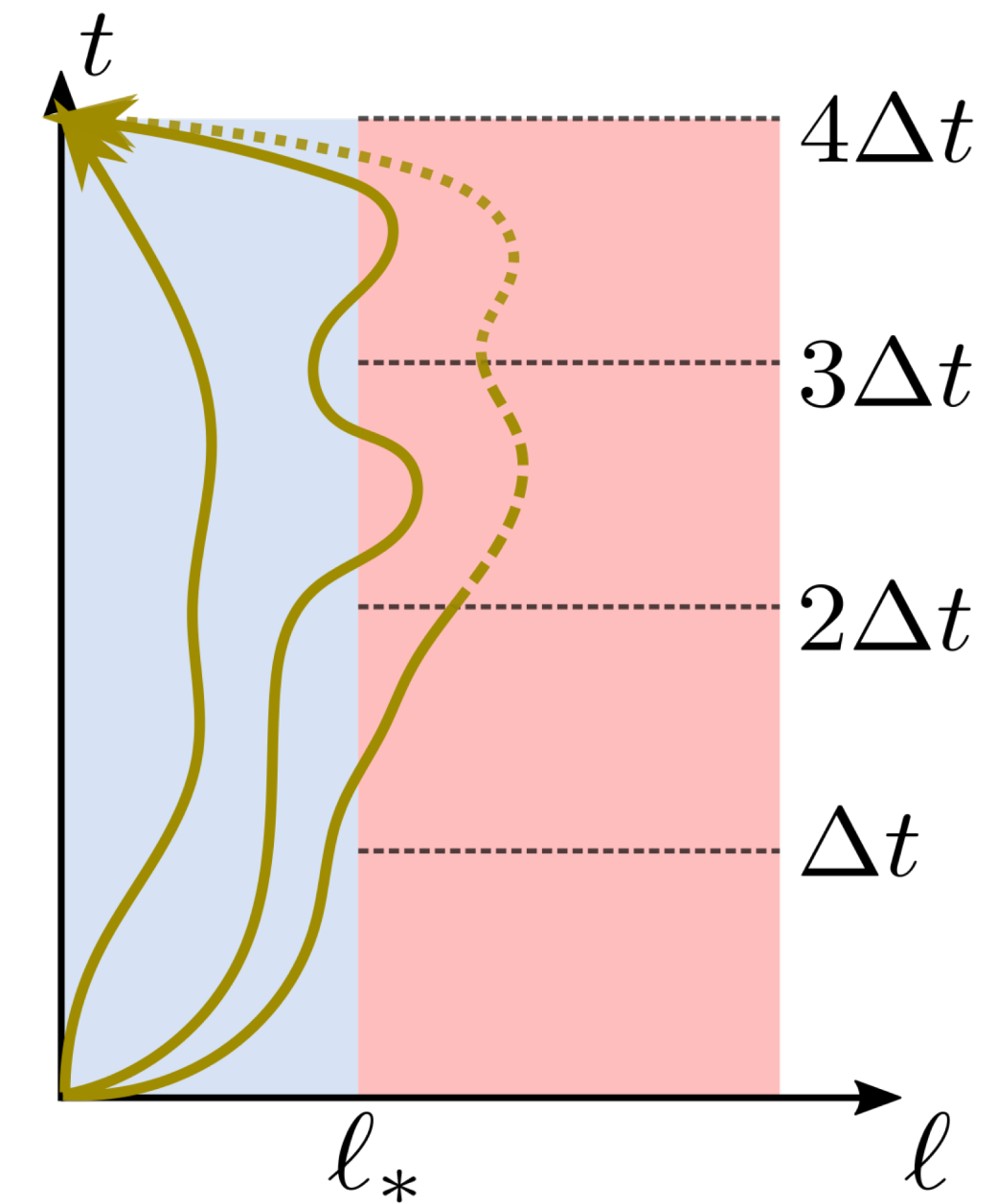
XXZ chain $H = \sum_{j=1}^L (X_j X_{j+1} + Y_j Y_{j+1} + \Delta Z_j X_{j+1})$



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 $n > \ell_*$.
- ▶ Errors induced by truncation (“backflow”) are exponentially suppressed in ℓ_*



[Rakovszky, von Keyserlingk, FP, PRB **105**, 075137 (2022)]

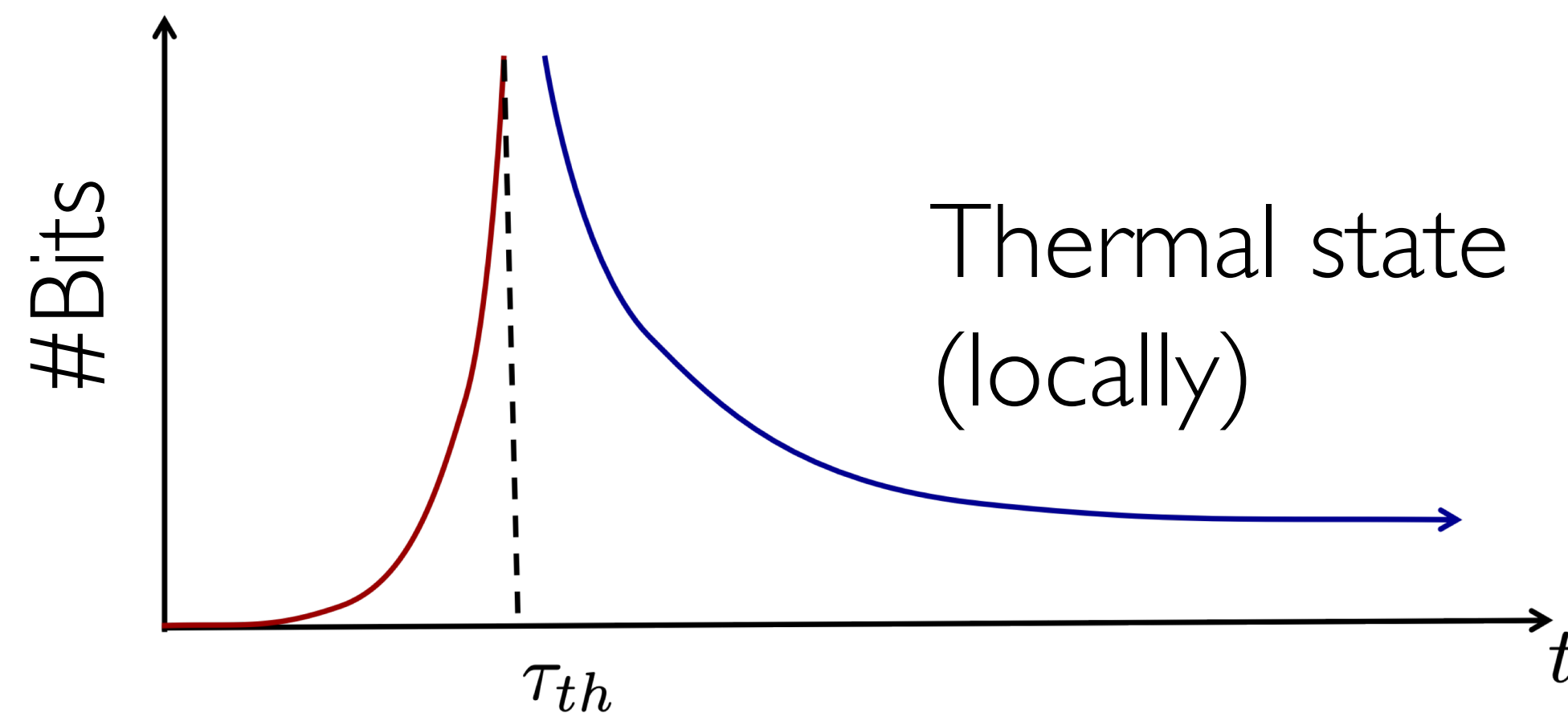
[von Keyserlingk, FP, Rakovszky, PRB **105**, 245101 (2022)]



Thank you!

Open questions and challenges

- (1) What is the optimal way to truncate entanglement without sacrificing crucial information on physical (local) observables?



- (2) What are the problems for which we do actually require a quantum computer to simulate physical (local) observables?