

Quantum many-body dynamics from neural networks with GPU acceleration



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1. Introduction

Solving quantum-mechanical problems is, in general, exponentially hard in the number of degrees of freedom N

$$|\psi\rangle = \sum_s \psi(s) |s\rangle$$

Quantum state \swarrow \nwarrow
 Complex amplitudes \uparrow Computational basis (e.g. spin configurations)

Key problem: full information requires storage of all amplitudes

Exponential growth of Hilbert space: $\#s \propto e^{CN}$

Number of amplitudes grows generally exponentially

- Possible solutions:
- efficient compression
 - reduction to effective simplified descriptions
 - ...

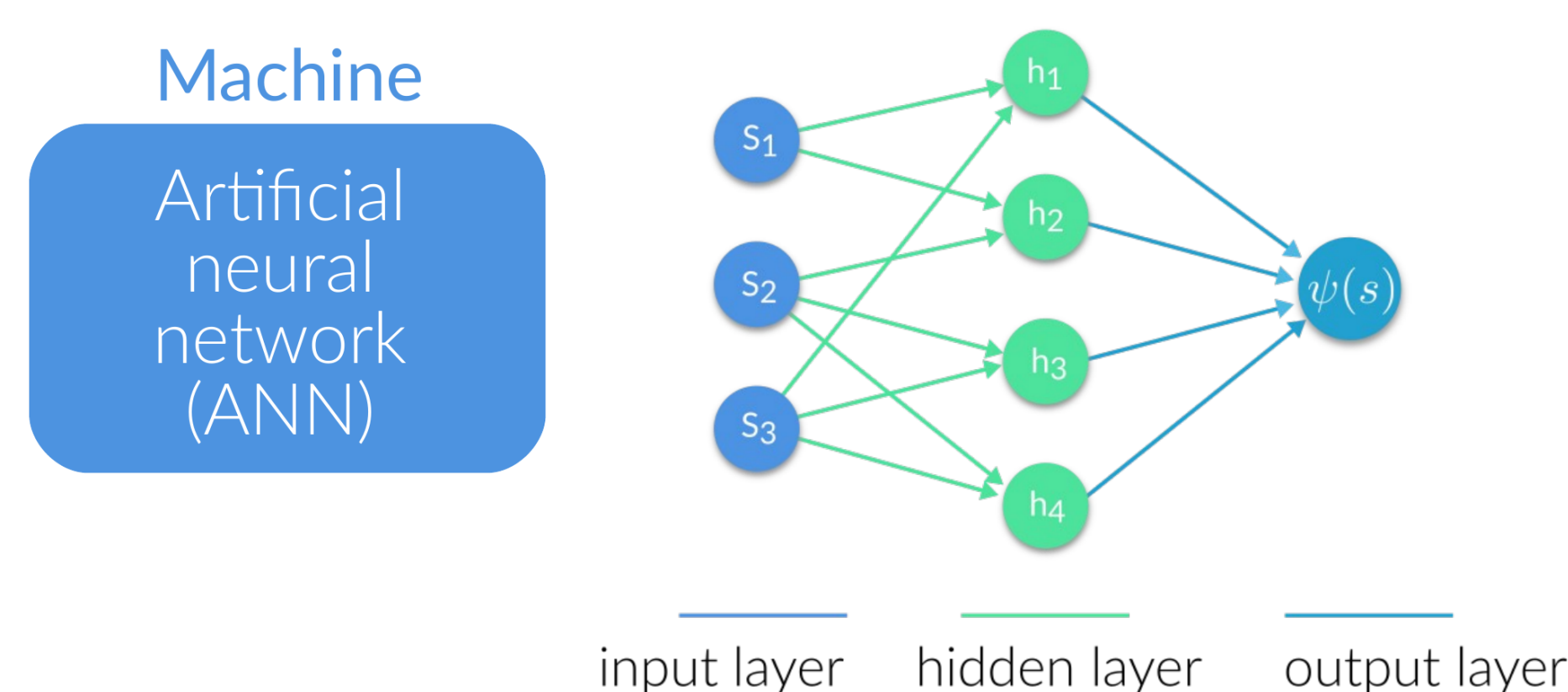
Challenge: No known (numerically exact) solution for nonequilibrium states in two and three spatial dimensions

2. Encoding quantum states in neural networks

Idea: "Don't store, generate on the fly"



- ▶ Sample from machine using Monte-Carlo techniques
- ▶ Avoids storage of amplitudes



Enabling idea: ANNs universal function approximators

- ▶ Any quantum state can be encoded in a (sufficiently large) ANN

Complexity of algorithm: poly(size of ANN, #DOF)

Carleo & Troyer, Science '17

Where's the problem then? Training....

Schmitt & Heyl, arXiv: 1912.08828

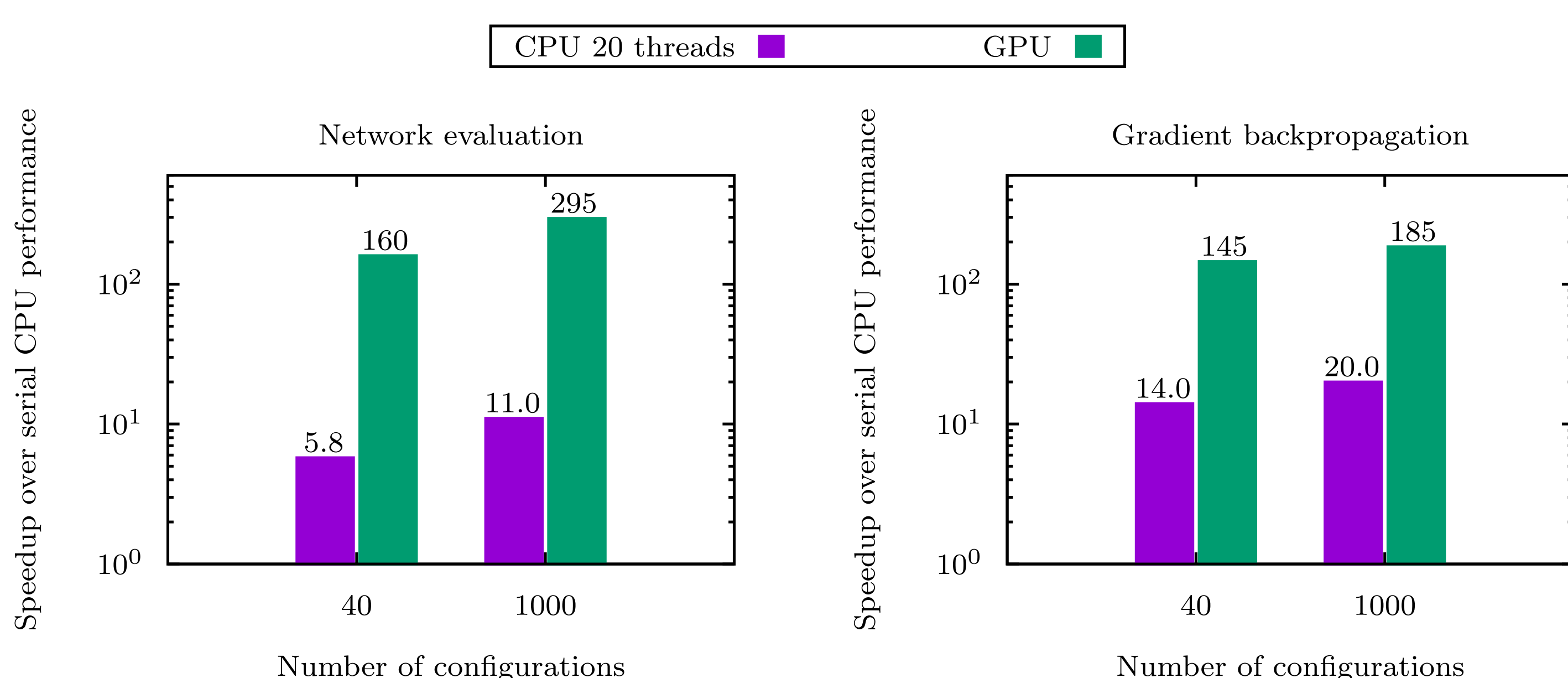
Quantum real-time evolution: time-dependent variational principle

Variational Monte-Carlo with ANN as variational wave function

3. Parallel implementation

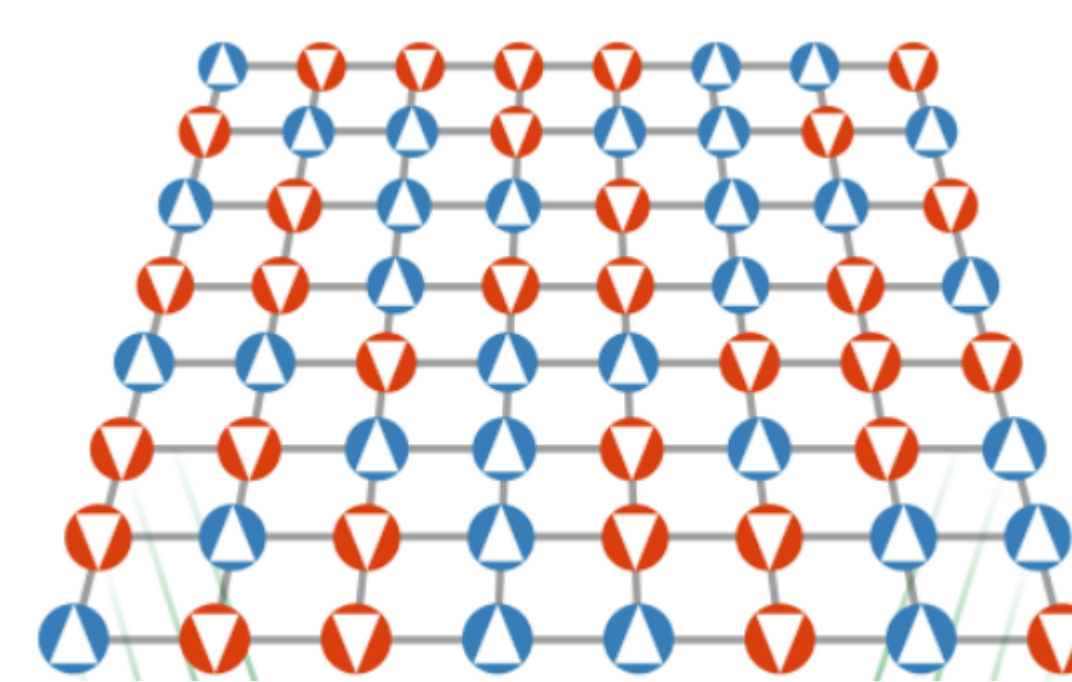
Multiple levels of parallelism:

- ▶ Monte-Carlo (trivially parallel) \longrightarrow MPI (multiple nodes)
- ▶ Network evaluation \longrightarrow CUDA (for GPUs)



3. Quantum dynamics in 2D

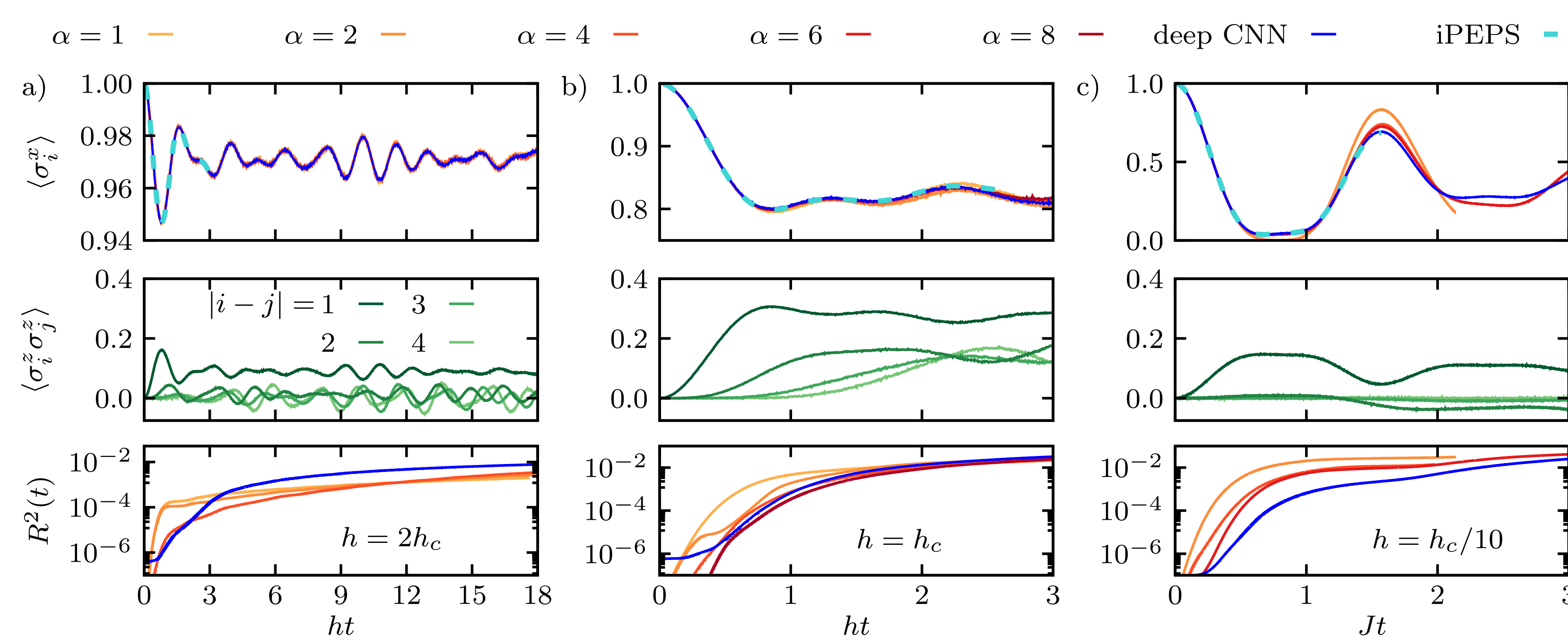
Schmitt & Heyl, arXiv: 1912.08828



Transverse-field Ising model on a square lattice

$$H = -J \sum_{\langle i,j \rangle} \sigma_i^z \sigma_j^z - h \sum_j \sigma_j^x$$

Quantum quench: initial condition $|\psi_0\rangle = |\rightarrow\rangle$



Agreement with iPEPS up to reached time scales
 iPEPS: numerically exact, longer times very hard

Main limitation: training & instabilities, not expressivity!

4. Quantum renormalization groups

- ▶ Fully many-body localized (MBL) systems: emergent constants of motion τ_i :

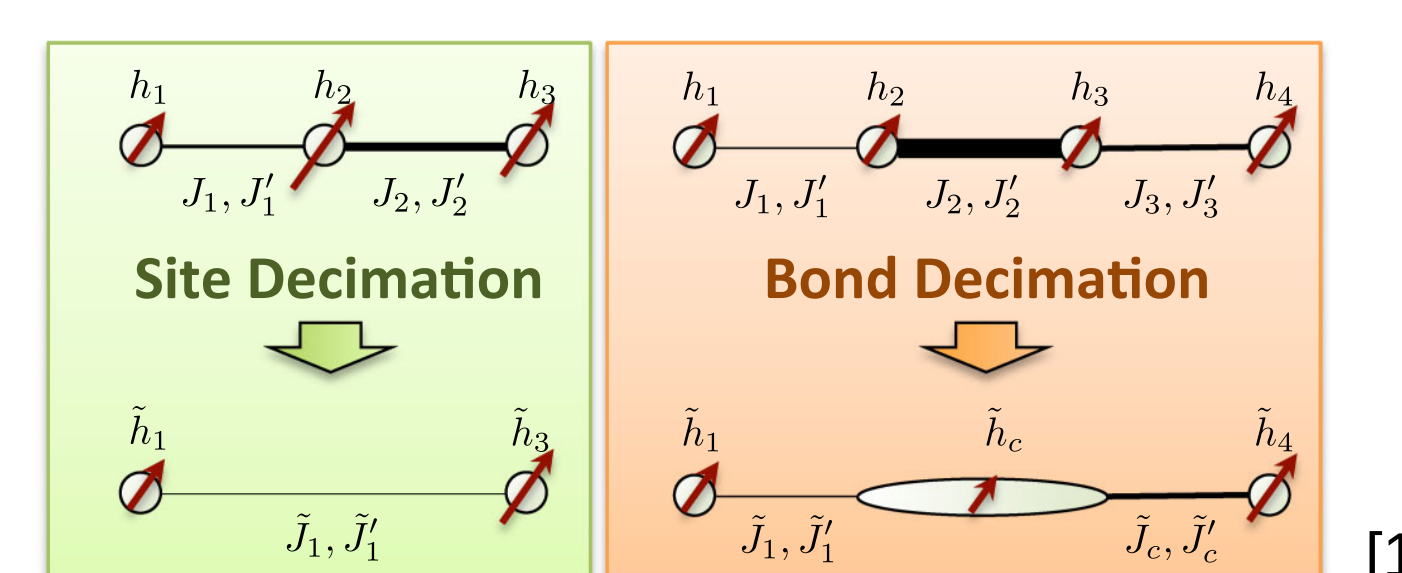
$$\tilde{H} = T^\dagger H T = \sum_i \alpha_i \tau_i + \sum_{ij} \beta_{ij} \tau_i \tau_j + \dots \quad T: \text{local unitary}$$

- ▶ **Idea:** use a quantum renormalization group (RG) transformation to obtain T .
- ▶ **Goal:** Describe non-equilibrium long-time dynamics in MBL systems.

Model: Non-integrable, disordered quantum Ising chain

$$H = \sum_{i=1}^L J_i \sigma_i^z \sigma_{i+1}^z + K_i \sigma_i^x \sigma_{i+1}^x + h_i \sigma_i^x, \quad J_i \in [0, J], K_i \in [0, 0.1 \cdot J], h_i \in [0, h]$$

Strong-disorder renormalization group



$$\tilde{H} = e^{-iS} H e^{iS} = \prod_m e^{-iS_m} H \prod_m e^{iS_m}$$

Schrieffer-Wolff (SW) transformation

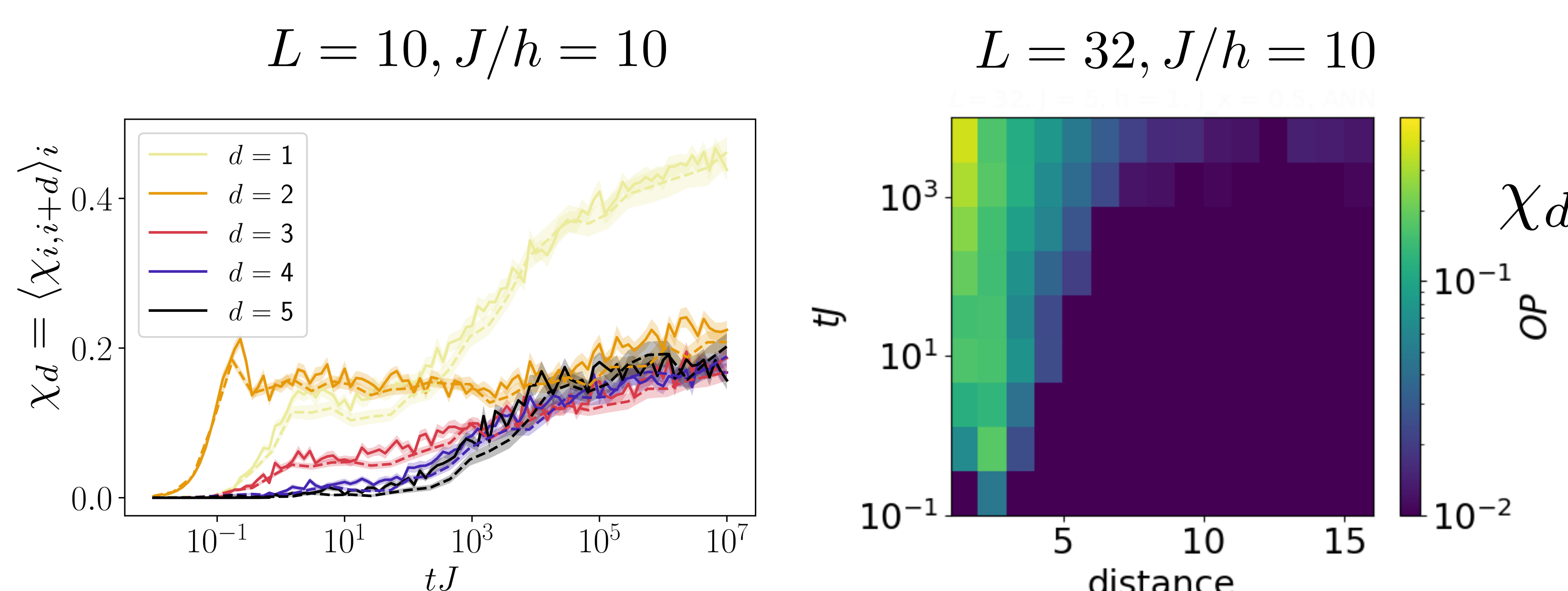
BUT: Taking into account the SW unitary for the calculation of observables is challenging \longrightarrow **encode the unitary in an artificial neural network (ANN).**

Idea: unitary transformation = time evolution

Thus: use same formalism as before

- ▶ **Quantum quench:** Time evolution starting from $|\psi_0\rangle = |\rightarrow\rangle$

- ▶ MBL-spin glass order parameter: $\chi_{i,j=i+d} = \sum_{n=1}^4 p_n^{ij} \langle \psi_n^{ij} | \sigma_i^z \sigma_j^z | \psi_n^{ij} \rangle^2$
 eigenvalues, eigenstates of reduced two-site density matrix



Comparison to exact diagonalization (dashed: RG + ANN, solid: ED)

Slow (logarithmic) buildup of long-range order