

# QUANTUM INSPIRED APPROACHES FOR SIMULATIONS OF DIFFERENTIAL EQUATIONS

Ryan J. J. Connor<sup>1,\*</sup> and Andrew J. Daley<sup>1</sup>

<sup>1</sup> Clarendon Laboratory, Department of Physics, University of Oxford, Oxford, UK.

\* ryan.connor@physics.ox.ac.uk

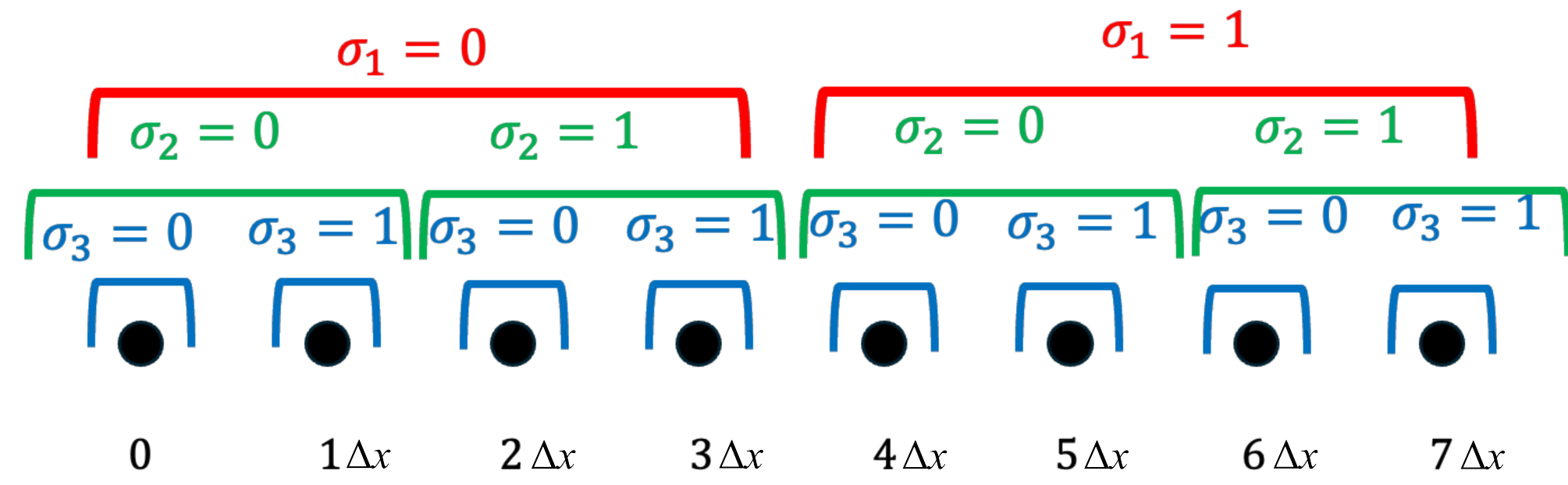


**Abstract:** Non-linear partial differential equations are ubiquitous across physics and engineering. For some applications, numerically accessible grid spacing can become a limiting factor leading to large computational resources. In recent years, quantum-inspired methods based on **tensor networks (TN)** have emerged as a promising route to address these issues on classical hardware. Tensor networks achieve physically motivated data compression without sacrificing length scales, which makes simulations possible on large spatial grids.

## 1. Encoding spatial grid into Tensor Network

Function  $f(x)$  defined on a uniform spatial grid given by  $x \in [0, 1]$  and  $\Delta x = 1/2^N$ . Can express in terms of a binary register,

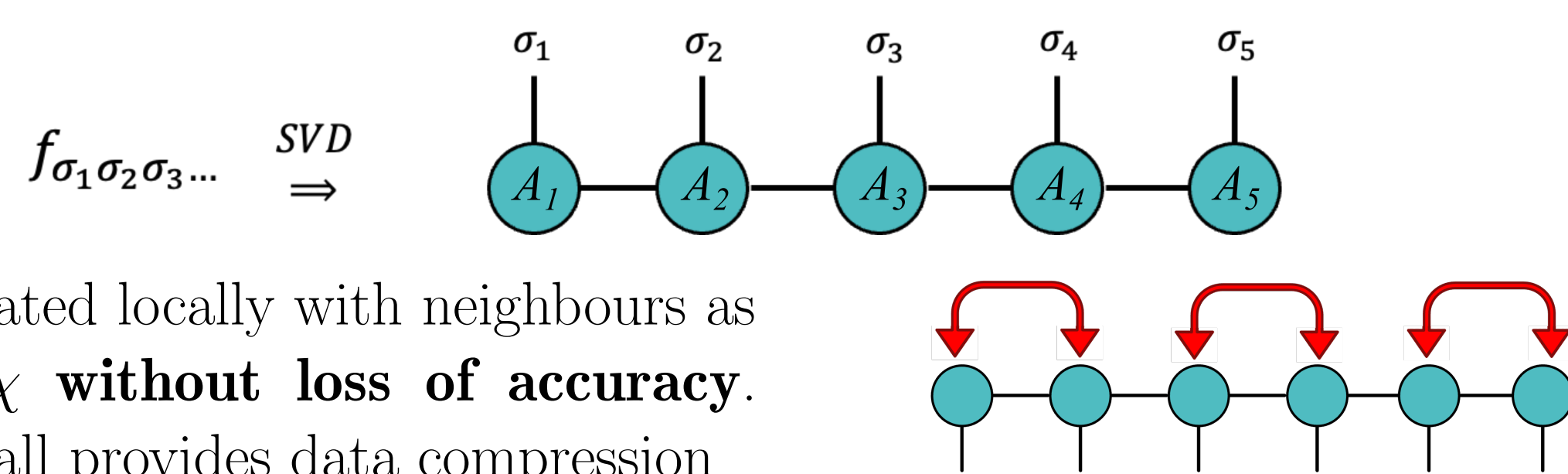
$$x = \frac{\sigma_1}{2^1} + \frac{\sigma_2}{2^2} + \frac{\sigma_3}{2^3} + \dots + \frac{\sigma_N}{2^N}.$$



Perform singular value decompositions to obtain a Matrix Product State (MPS) [1],

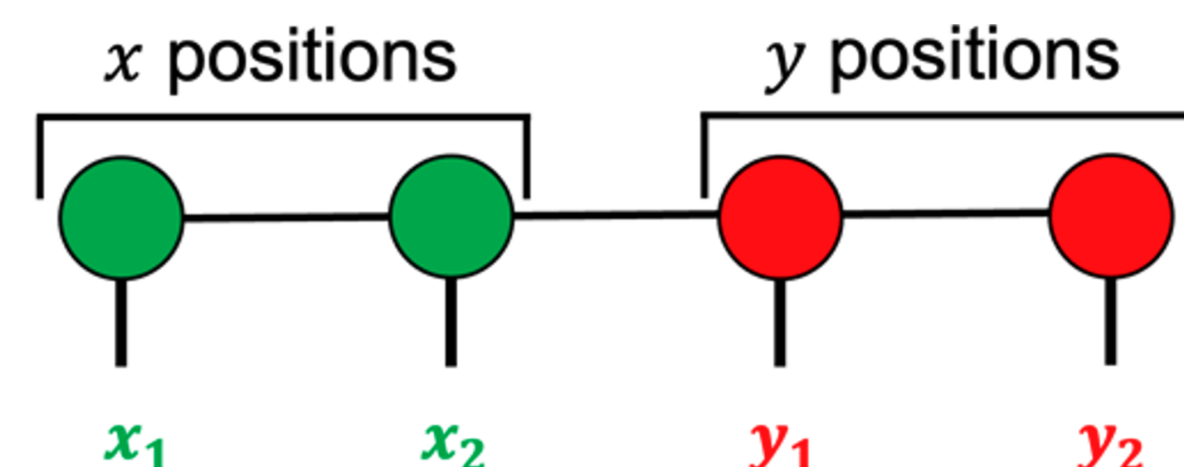
$$f(x) = f_{\sigma_1, \sigma_2, \dots} = A_1^{\sigma_1} A_2^{\sigma_2} A_3^{\sigma_3} \dots A_N^{\sigma_N}.$$

Define  $\chi$  as number of rows/cols of each tensor  $A^n$ , known as bond dimension. Each tensor represented as circle with lines emerging. Connecting bonds carry correlations.



Desire tensors correlated locally with neighbours as all to **restrict**  $\chi$  **without loss of accuracy**. Limiting  $\chi$  to be small provides data compression

Can extend the MPS representation to 2D and beyond. How data is encoded into MPS is important but non-unique. Often choose to encode each dimension sequentially.



## 2. Finite Differences

Given a function in MPS form, can construct Matrix Product Operators (MPOs) to perform differentiation. Express the function  $f$  in a quantum register  $|f\rangle$  with  $N$  qubits  $|\sigma_1 \sigma_2 \dots \sigma_N\rangle$ ,

$$|f\rangle = \sum_{\sigma_1, \sigma_2, \dots} f_{\sigma_1, \sigma_2, \dots, \sigma_N} |\sigma_1, \sigma_2, \dots, \sigma_N\rangle.$$

The differential operator can be approximated using raising/lowering operators, which can then be written as an MPO with  $\chi = 3$  [2],

$$\hat{\partial}_x \approx \frac{\hat{S}^+ - \hat{S}^-}{2\Delta x} = \frac{1}{2\Delta x} \begin{pmatrix} s_1^+ & s_1^- & I_1 \end{pmatrix} \left( \begin{smallmatrix} s_n^- & s_n^+ \\ s_n^+ & s_n^- & I_n \end{smallmatrix} \right)^{\otimes N-2} \begin{pmatrix} s_N^- \\ -s_N^+ \\ s_N^+ - s_N^- \end{pmatrix}.$$

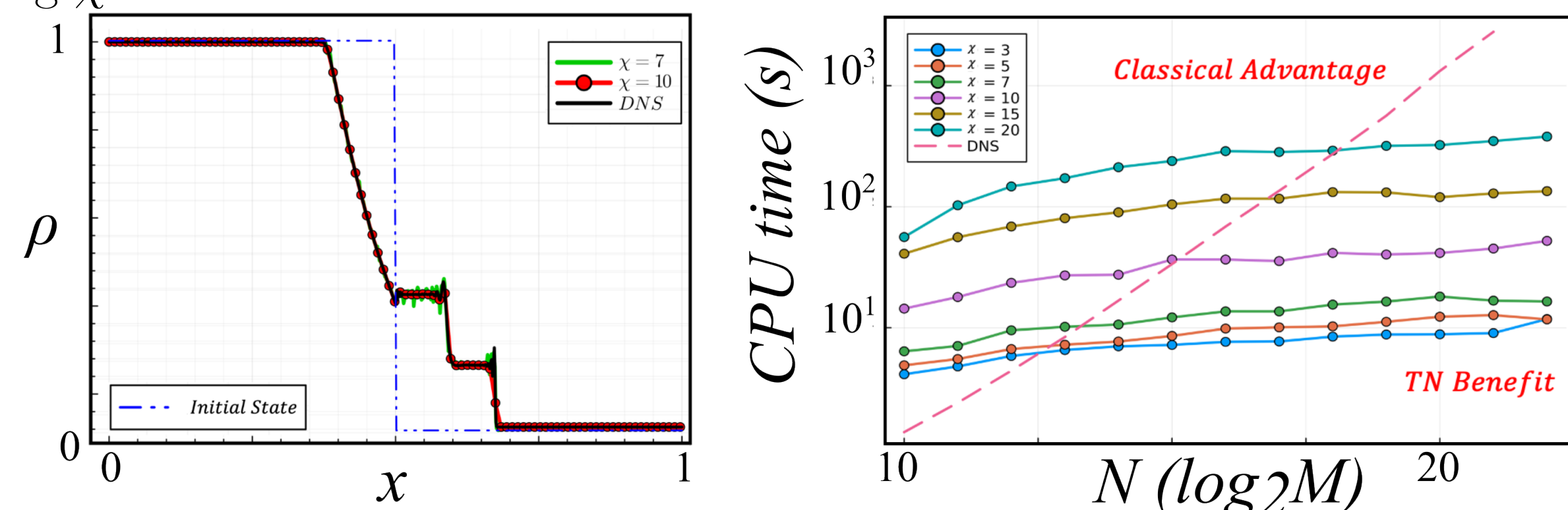
$S^\pm$  are global raising/lowering operators on the binary string, and  $s_n^\pm$  are local raising/lowering operators on the  $n$ th qubit.

## 3. Compressible flows

Compressible flows governed by Euler equations. Each variable encoded as separate MPS with bond dimension  $\chi$ . Time evolve via finite difference schemes and Macormack evolution.

$$\begin{aligned} \frac{\partial \rho}{\partial t} + \nabla \cdot (\rho \mathbf{u}) &= 0, \\ \frac{\partial \rho \mathbf{u}}{\partial t} + \nabla \cdot [\rho \mathbf{u} \mathbf{u}^T + p] &= 0, \\ \frac{\partial E}{\partial t} + \nabla \cdot [(E + p) \mathbf{u}] &= 0. \end{aligned}$$

Standard test case of Sod Shock [3]. Initial (dashed blue) discontinuity left to evolve, leading to shock formation. Only small  $\chi$  required to reproduce dynamics, leading to data compression. Truncating  $\chi$  too small leads to errors in simulation.



True benefit comes via scaling. Conventional solvers scale with grid points ( $M$ ) i.e.  $\sim M$ . TN solvers scale logarithmically with  $M$  and polynomially in  $\chi$ , i.e.  $\sim \chi^\alpha \log_2 M$ .

Can allow TNs to simulate dense grids which would be infeasible with direct simulations, provided  $\chi$  can be limited without large errors. This appears to be the case for many PDEs!

## References

[1] Ulrich Schollwöck. In: *Annals of Physics* 326.1 (2011). [2] Martin Kiffner and Dieter Jaksch. 2023. arXiv: 2303.03010. [3] Gary A Sod. In: *J Comp Phys* 27.1 (1978). [4] Woo Jin Kwon et al. In: *Phys. Rev. Lett.* 117 (24 2016).

## 4. Quantised Vortices

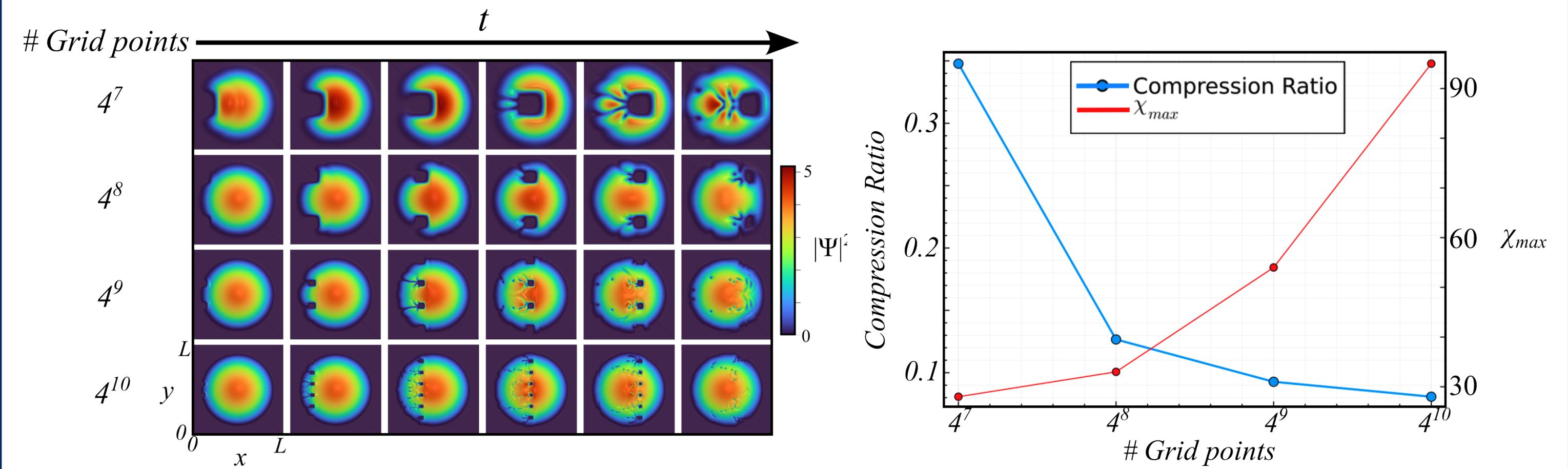
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Dynamics of Bose-Einstein condensate (BEC) governed by the Gross-Pitaevskii Equation,

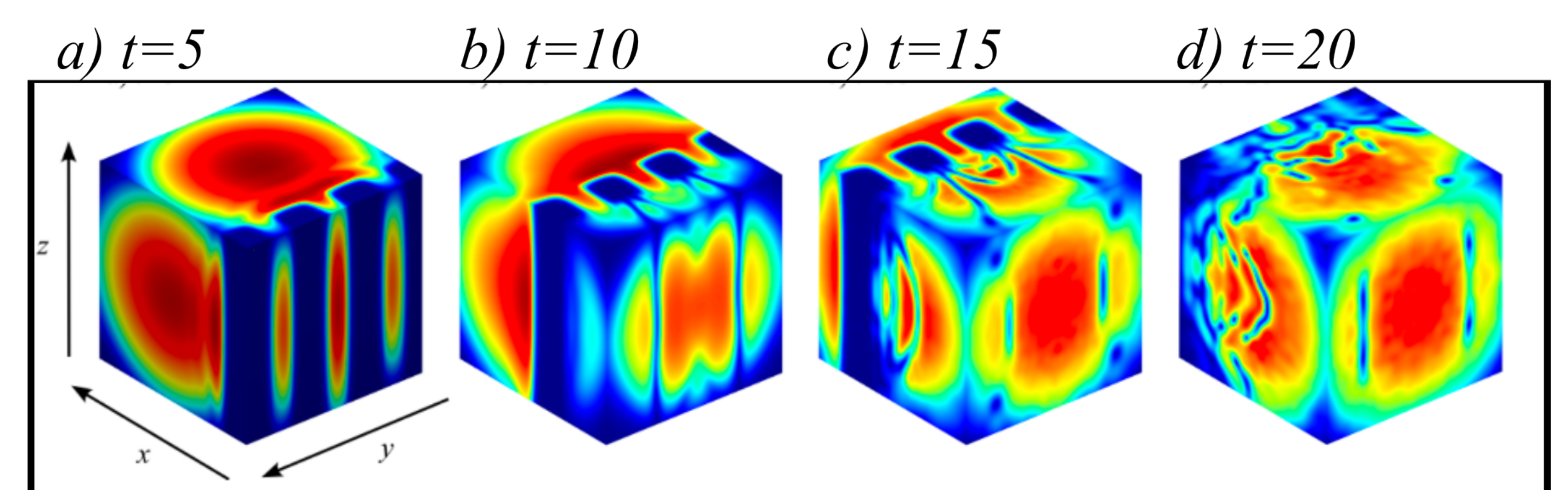
$$i\hbar \frac{\partial \Psi}{\partial t} = (1 - i\gamma) \left[ -\frac{\hbar^2}{2m} \nabla^2 + V(x, t) - \mu + g|\Psi|^2 \right] \Psi.$$



Apply an external potential  $V(x, t)$ , which sweeps through and generates vortices behind [4]. Perform MPS simulations for increasing grid density, resolving finer and finer vortices, making use of adaptive  $\chi$ , based on desired error.



Plot resulting  $\chi$  as function of grid-points for fixed error ( $10^{-12}$ ). Observe polynomial increase in  $\chi$  as function of grid density. Defined compression as ratio of number of terms in MPS simulation to that of a direct simulation. **Huge degrees of compression available for large, dense spatial grids.** Less than 10% resources required for  $2^{10} \times 2^{10}$  grid.



Can extend to 3D simulations. Shown dynamics of  $2^7 \times 2^7 \times 2^7$  spatial grid with fixed  $\chi = 64$ . Corresponds to compression to **4% numerical parameters** compared to direct simulations.

## 5. Plasma Physics

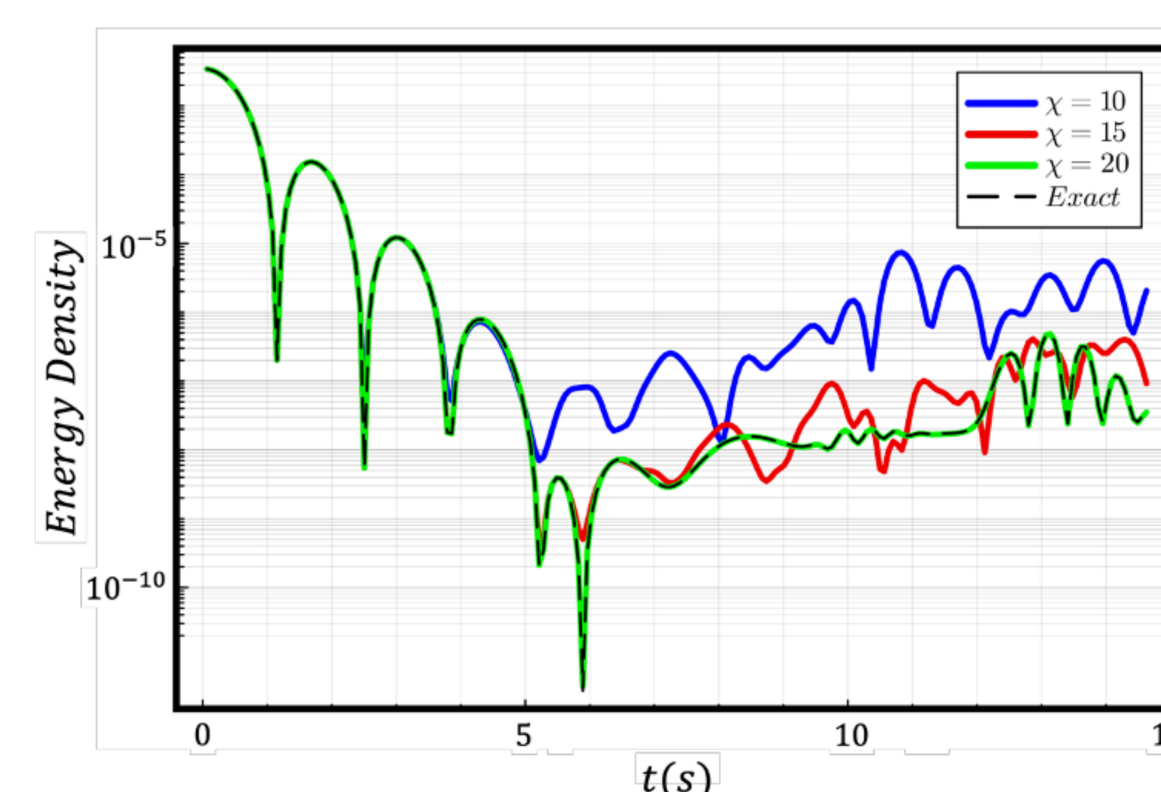
arXiv:2512.15924

Plasmas governed by Vlasov-Maxwell or Magnetohydrodynamics,

### Vlasov-Maxwell

$$\begin{aligned} \frac{\partial f_s(\mathbf{x}, \mathbf{v}, t)}{\partial t} + \mathbf{v} \cdot \nabla_{\mathbf{r}} f_s + \frac{q_s}{m_s} \mathbf{E} \cdot \nabla_{\mathbf{v}} f_s &= 0, \\ \frac{\partial \mathbf{E}}{\partial t} &= c^2 \nabla \times \mathbf{B} - \frac{1}{\epsilon_0} \mathbf{J}, \\ \frac{\partial \mathbf{B}}{\partial t} &= -\nabla \times \mathbf{E}. \end{aligned}$$

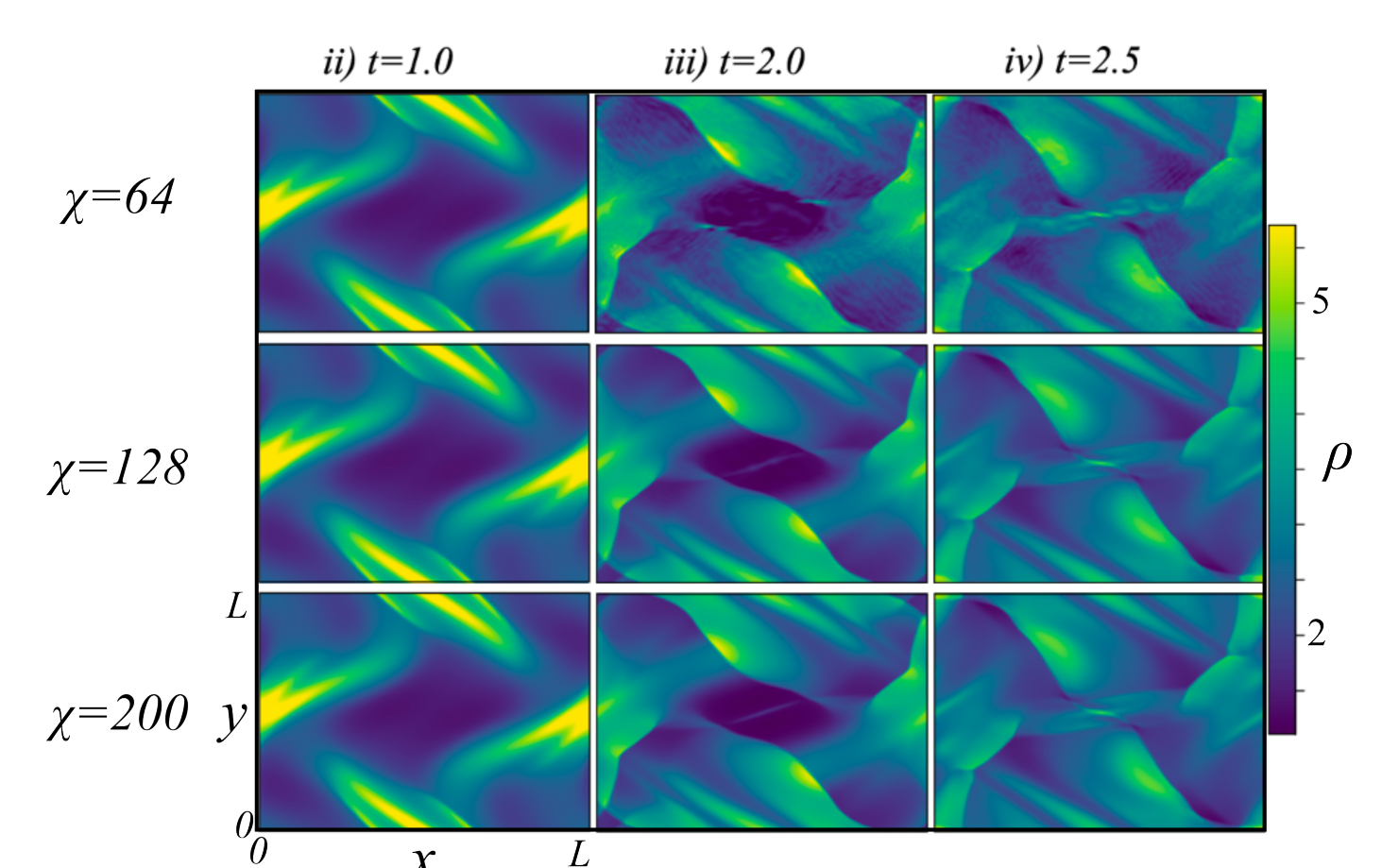
VM models phase space distribution  $f(\mathbf{x}, \mathbf{v}, t)$ . Must encode spatial **and** velocity dimensions into MPS. Electric ( $\mathbf{E}$ ) and Magnetic ( $\mathbf{B}$ ) fields represented as MPS and evolved via Maxwells equations. VM leads to fewer equations, but higher dimensionality, (more tensors per MPS).



### MHD

$$\begin{aligned} \frac{\partial \rho}{\partial t} &= -\nabla \cdot (\rho \mathbf{u}), \\ \frac{\partial (\rho \mathbf{u})}{\partial t} &= -\nabla \cdot [\rho \mathbf{u} \mathbf{u}^T + (p + \frac{1}{2} B^2) I_{3 \times 3} - \mathbf{B} \mathbf{B}^T], \\ \frac{\partial \mathbf{E}}{\partial t} &= -\nabla \cdot \left[ \left( \frac{\gamma}{\gamma-1} p + \frac{1}{2} \rho u^2 \right) \mathbf{u} - (\mathbf{u} \times \mathbf{B}) \times \mathbf{B} \right], \\ \frac{\partial \mathbf{B}}{\partial t} &= \nabla \times (\mathbf{u} \times \mathbf{B}). \end{aligned}$$

MHD derived from VM. Used for stellar formation and large scale turbulence modelling. MHD treats plasma as a compressible fluid in real space only, with a fluid density  $\rho(\mathbf{x})$  and velocity  $\mathbf{u}(\mathbf{x})$ . Each of dynamical variables encoded as MPS, giving more simultaneous equations, but lower dimensionality.



Across both VM and MHD formalisms, MPSs can efficiently capture underlying physics and recreate dynamics with **small bond dimensions (high compression ratios)**

## Conclusions + Future directions

- TNs can be applied to **wide range of PDEs**, most notably for fluid and plasma dynamics
- Data compression via **removing correlations as opposed to discarding length scales**
- Further development underway to allow TNs to simulate in spatially dense regimes impossible with direct simulations
- Can extend to **curvilinear grids**. Research ongoing for further complex topologies and enhancement of solvers beyond finite difference.
- Alternative TN geometries may present further opportunities for data compression.