

# Message Passing is all you need!

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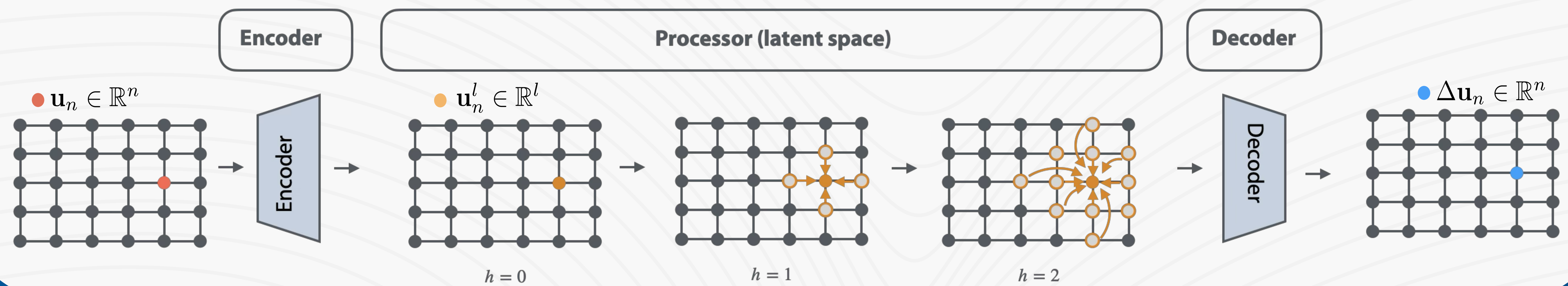
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## Graph Neural Networks Simulating PDEs with AI?

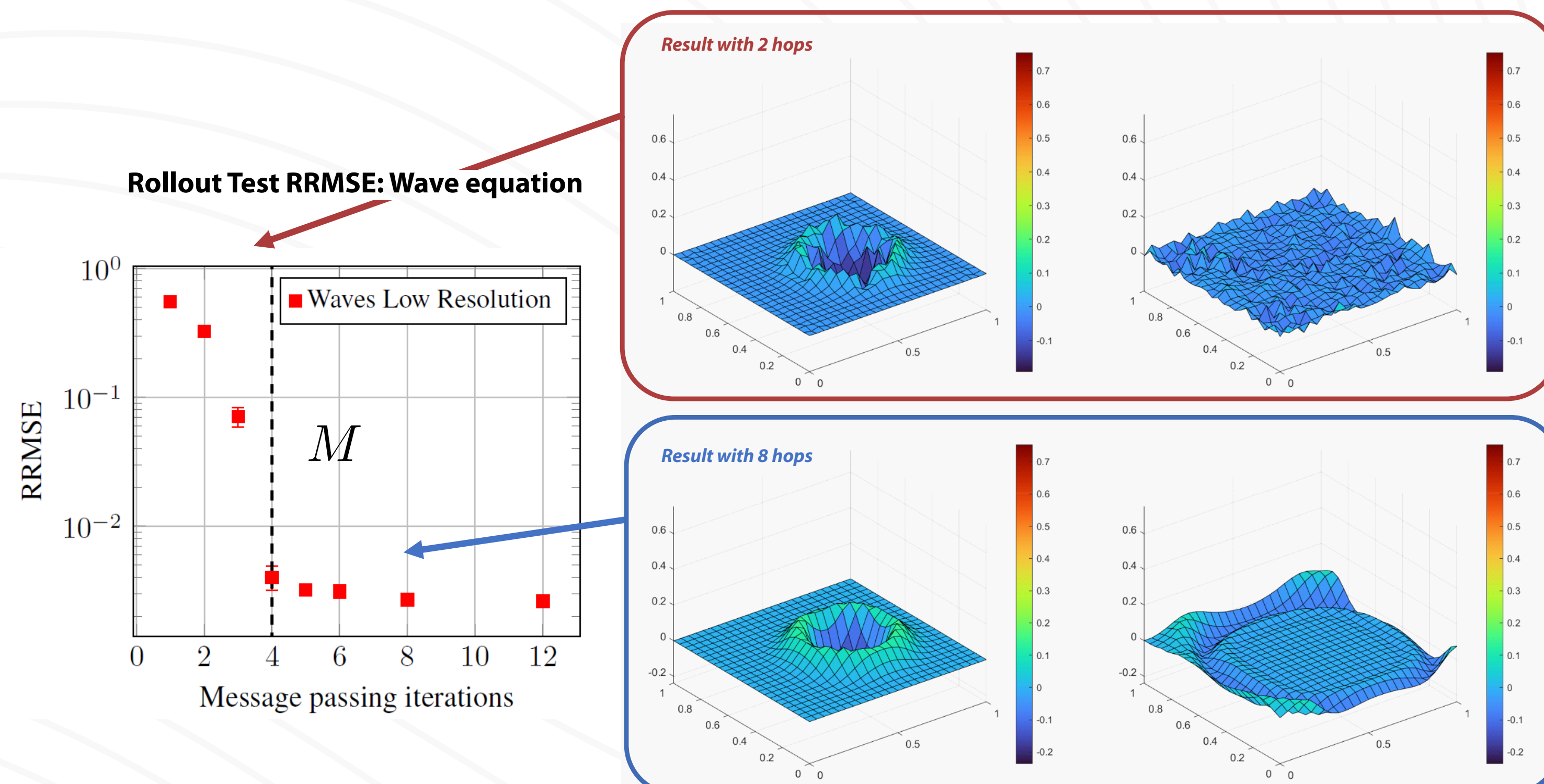
Explicit Scheme:  $\Delta \mathbf{u}_n = \text{GNN}(\mathbf{u}_n, \theta)$   
 $\mathbf{u}_{n+1} = \mathbf{u}_n + \Delta \mathbf{u}_n,$

Node processor - Message aggregation:  $\xi_i^{h+1} = \varphi \left( \xi_i^h, \sum_{j \in \mathcal{N}(i)} \xi_j^h, \theta \right)$



## How many message passes must we do?

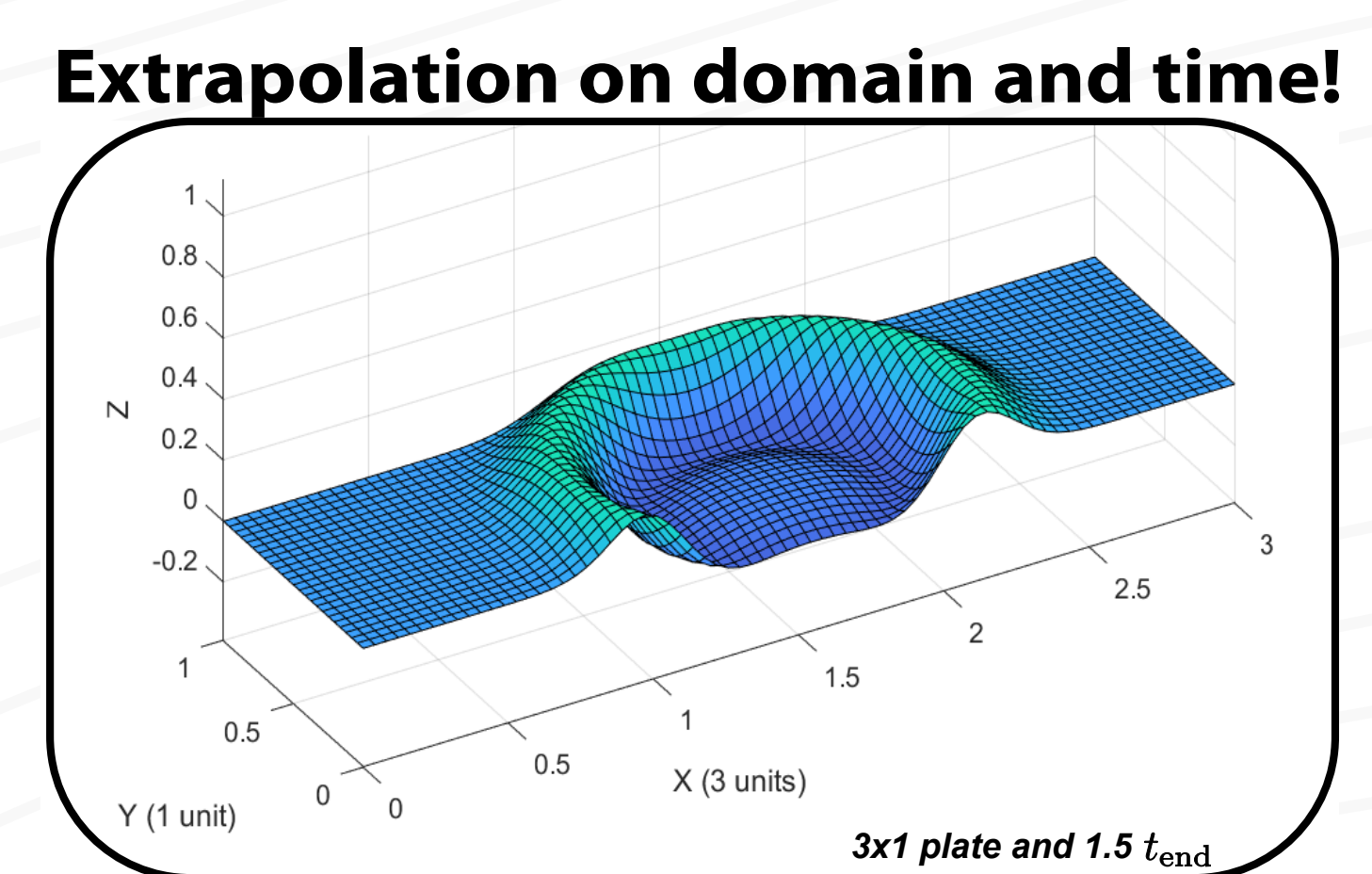
Phenomena governed by hyperbolic equations



$$\frac{\partial^2 u}{\partial t^2} = c^2 \left( \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right) = c^2 \nabla^2 u, \quad M > \sqrt{2} \frac{c \Delta t}{\Delta x}$$

! If we model an hyperbolic system with a GNN the wave propagation cannot exceed the message transmission distance in any time increment. This limit is designated as  $M$ .

But once satisfied...



Phenomena governed by parabolic and elliptic PDEs

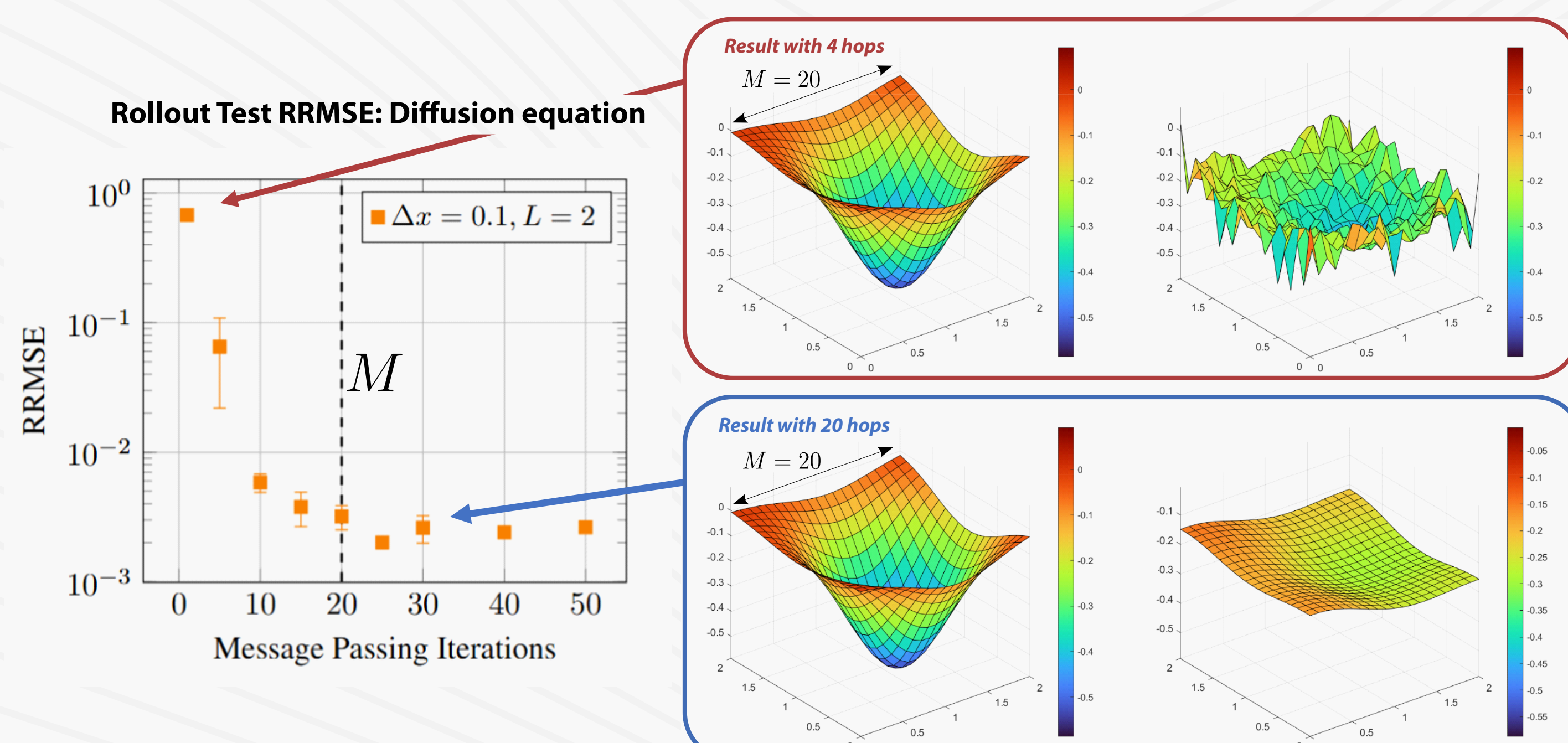
! Convergence requires **global information propagation** across the domain.

$$M = \frac{L}{\Delta x}$$

where  $L$  denotes the size of the domain,  $\Delta x$  is the spatial discretization.

Thermal Diffusion Equation

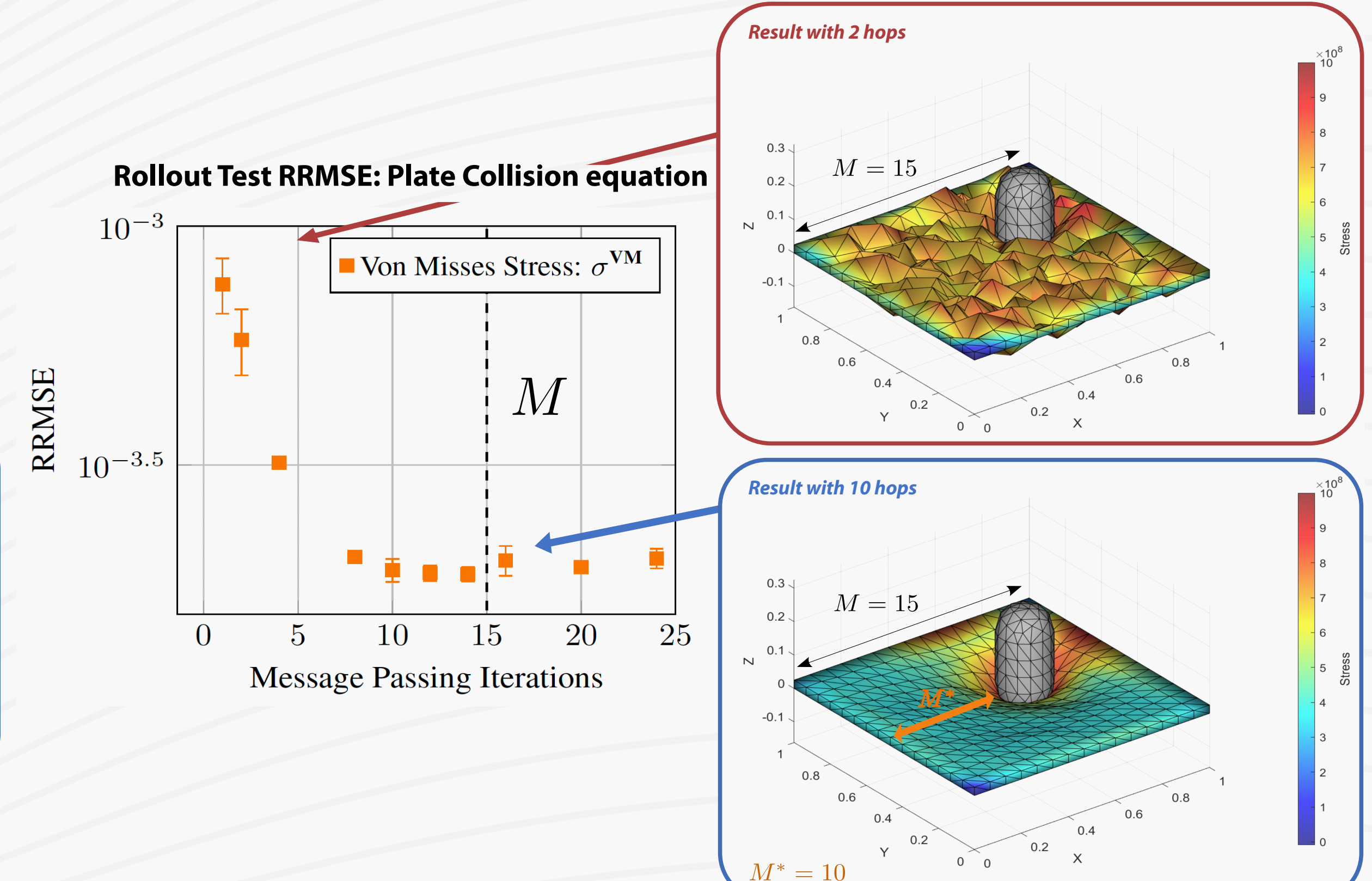
$$\frac{\partial u}{\partial t} = \alpha \nabla^2 u \quad \text{in } \Omega \times (0, T]$$



Incremental Forming

$$\nabla_0 P + f_0 = 0,$$

regulated by a yield function  $\mathbf{F} = \mathbf{F}^{el} \cdot \mathbf{F}^{pl}$



## References

- [1] Johannes Brandstetter, Daniel Worrall, and Max Welling. Message passing neural pde solvers. arXiv preprint arXiv:2202.03376, 2022.
- [2] Tobias Pfaff, Meire Fortunato, Alvaro Sanchez-Gonzalez, and Peter Battaglia. Learning mesh-based simulation with graph networks. In International Conference on Learning Representations, 2021.
- [3] Uri Alon and Eran Yahav. On the bottleneck of graph neural networks and its practical implications. arXiv preprint arXiv:2006.05205, 2020.