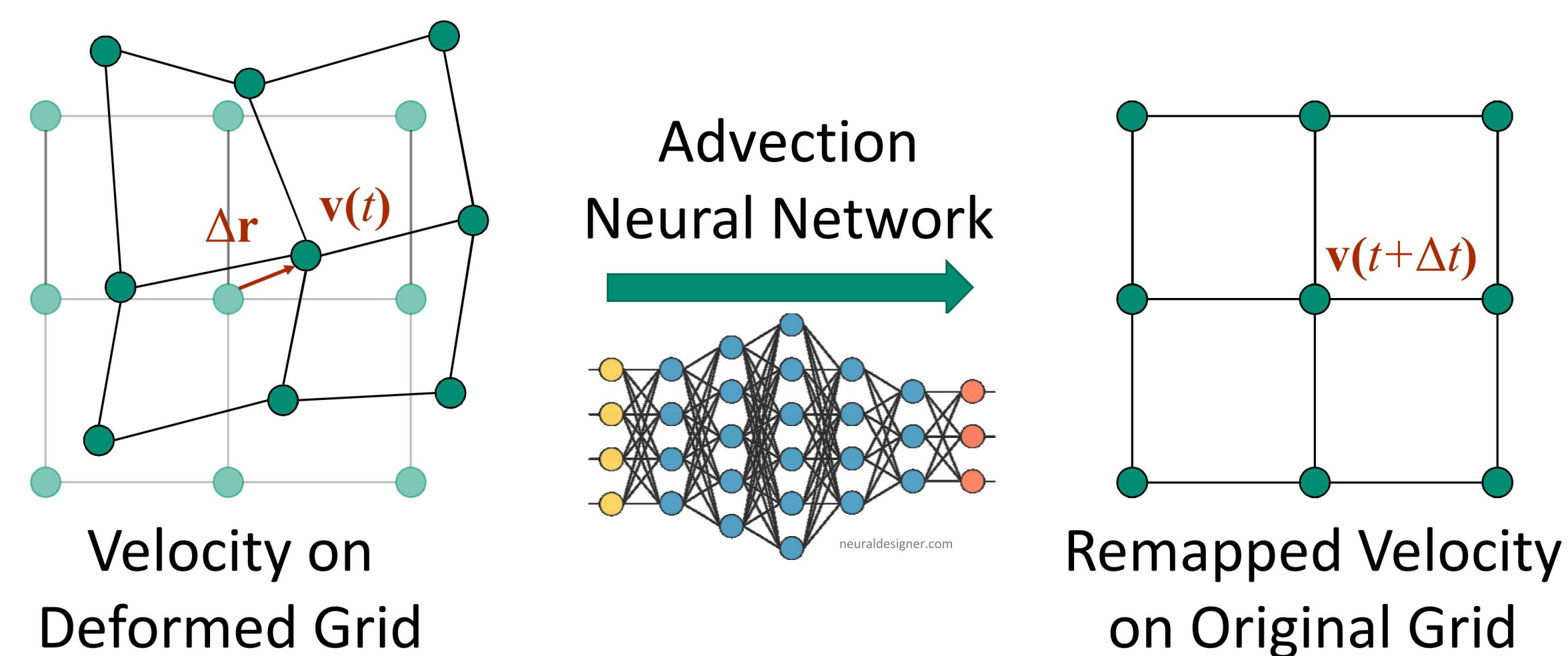


# Deep Learning the Advection in Eulerian Hydrocodes

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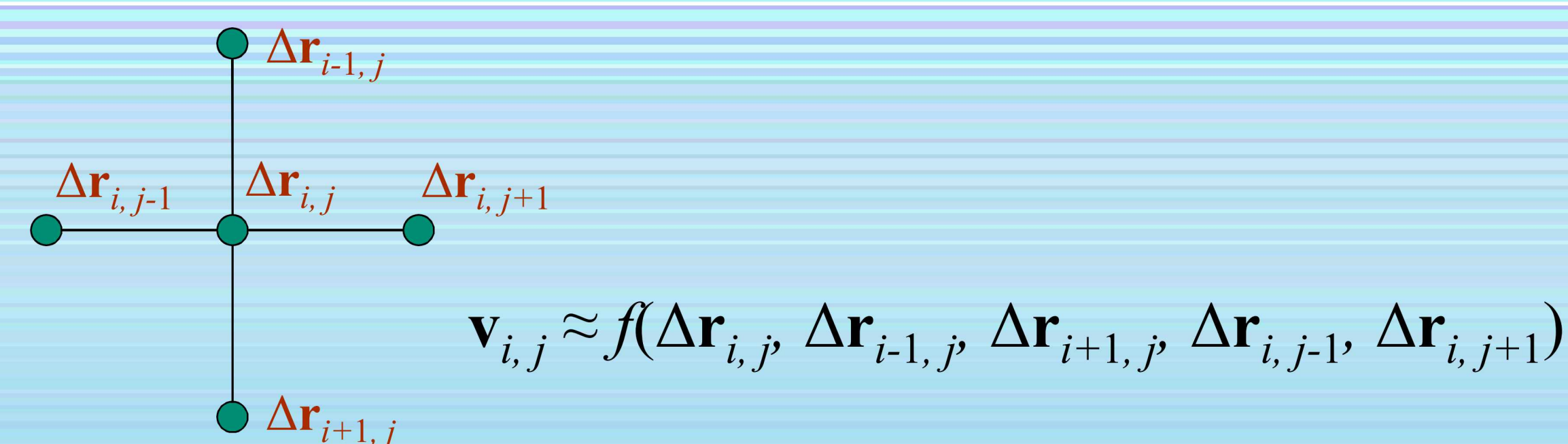
## Problem

- Can we **substitute** traditional components of a simulation algorithm with **data-driven models**?



- 2 steps of hydrocode**
  - Lagrangian step: Deform mesh
  - Remap/Advection Step: Remap nodal velocities back onto original mesh
- Objective is to approximate the remap step using a neural network**

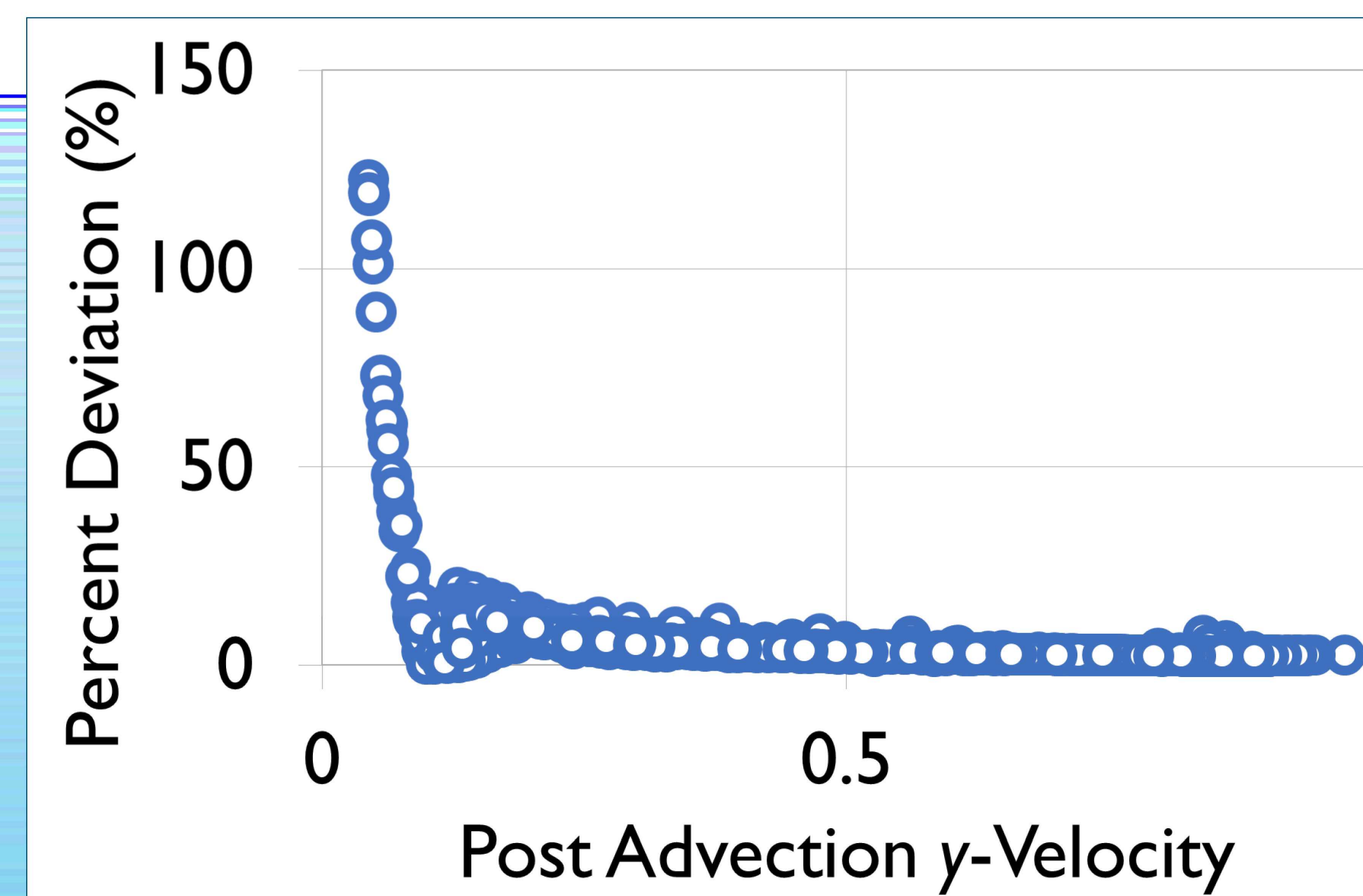
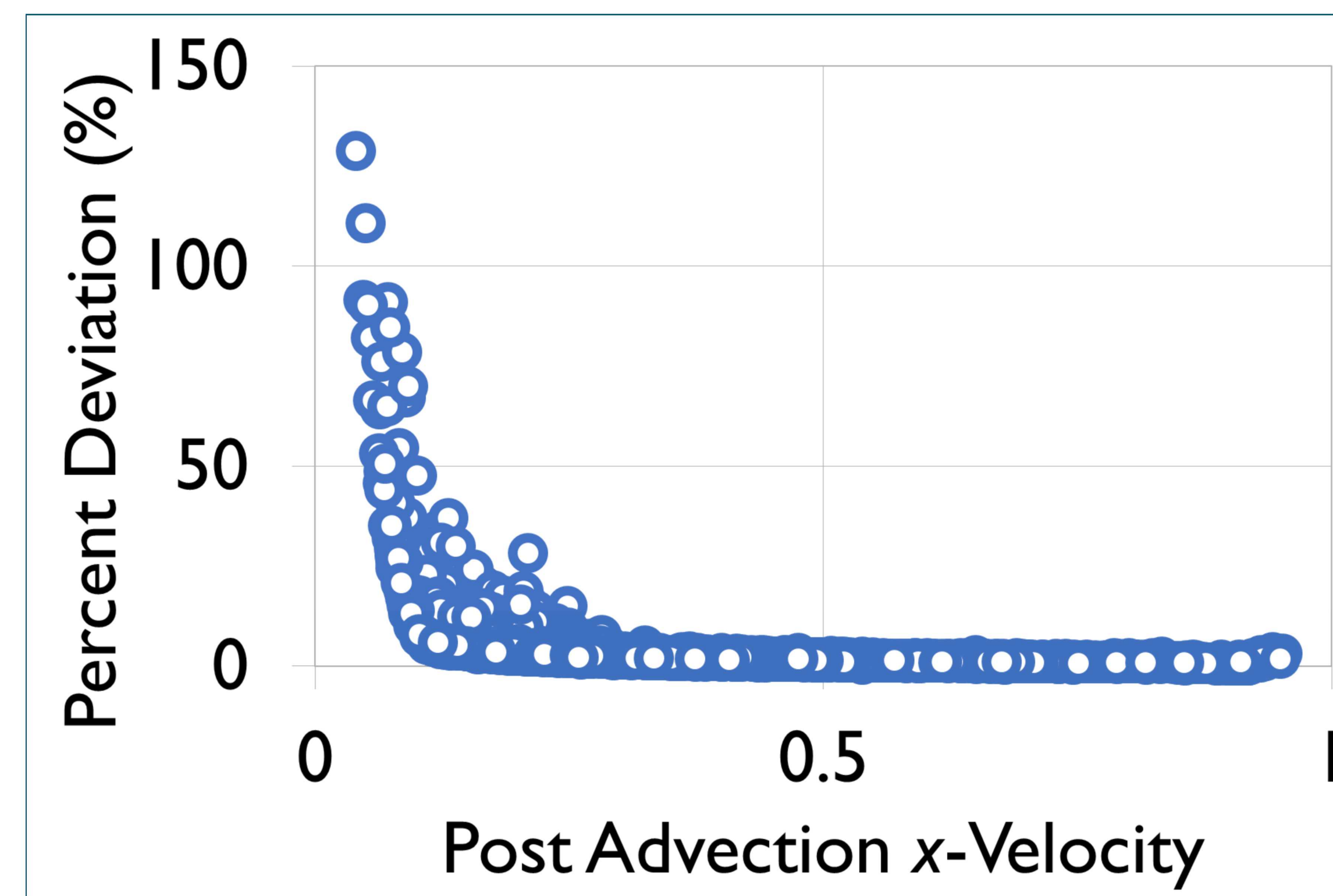
## Training Data



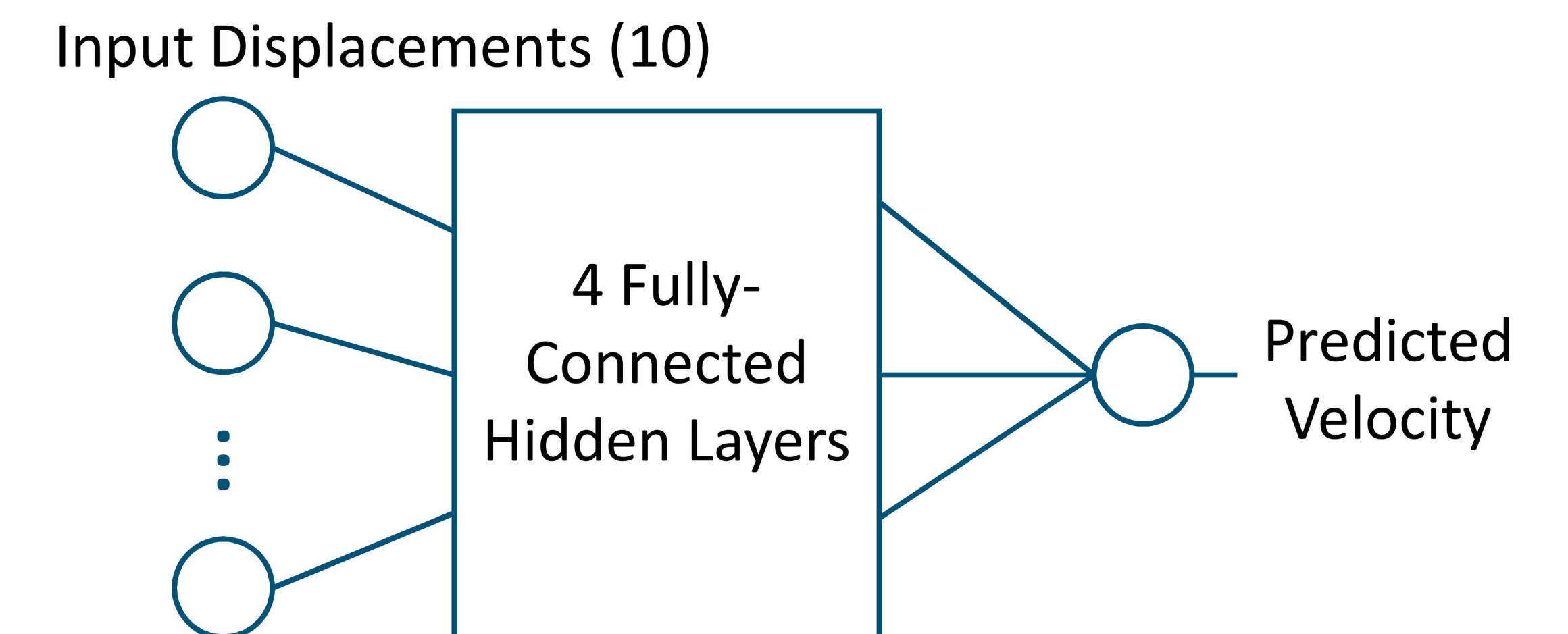
- ~100,000 data points** of post advection velocity vs. displacement at nodes and immediate neighbors
- Randomly initialized, normally distributed velocities on 2D grid (normalized between 0 and 1)

## Results

- Average predicted velocities **within 4%**
- Larger deviations near zero velocities due to divide-by-zero effect



## Network Architecture



- Multi-layer perceptron, Leaky ReLU activation
- Smooth L1 loss optimized using SGD
- Two separate networks trained to predict  $x$  and  $y$  velocity components

## Significance & Outlook

- This exploratory work shows promise in accurately predicting advection velocity
  - More advanced models and/or more data will likely improve accuracy
  - Other algorithms in computational solvers should be explored
- An advanced DNN-remap algorithm trained on high resolution simulations or exact solutions may be possible

## Acknowledgment

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