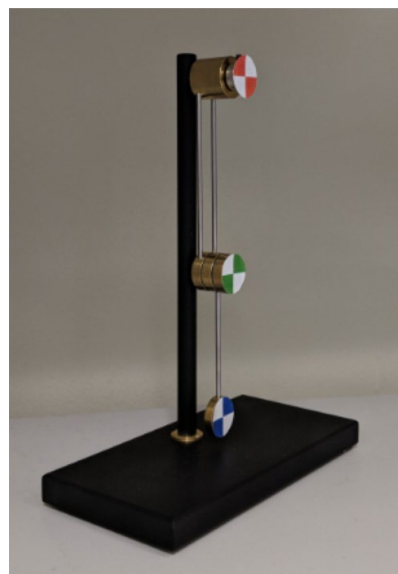


Sparse-Input Neural Network Augmentations for Differentiable Simulators

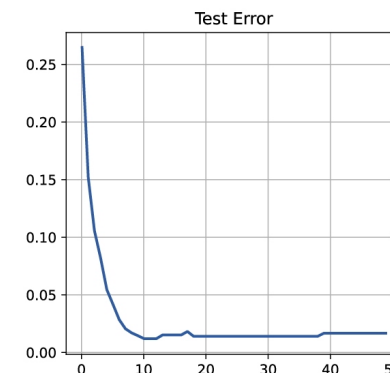
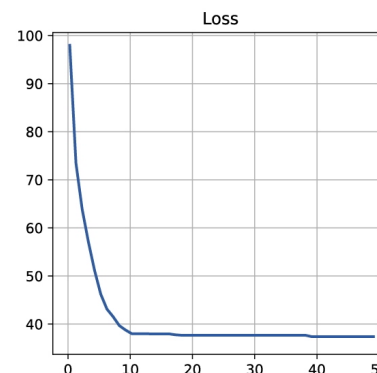
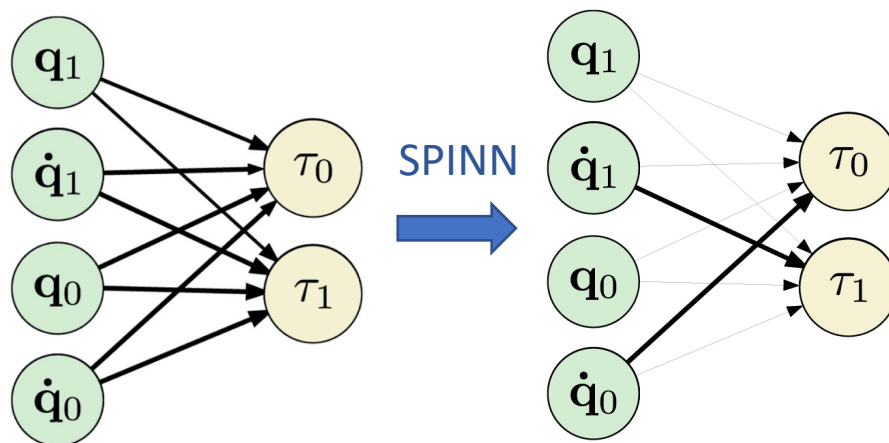
Eric Heiden, David Millard, Erwin Coumans, Gaurav S. Sukhatme

- Insert neural network connections at any place inside a differentiable rigid-body physics engine to learn unmodeled effects from data
- Simplify neural augmentations via **sparse-group lasso** loss \mathcal{L} used for training **Sparse-Input Neural Networks (SPINN)**

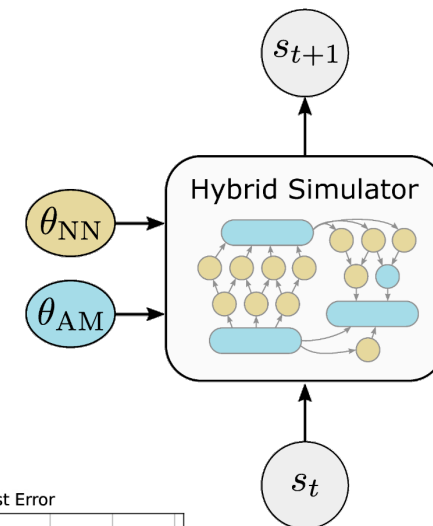


IBM Dataset [Asseman 2018]

Experiment: discover friction force (τ) dependence on joint velocity $\dot{\mathbf{q}}$



Sparse-group lasso loss:
$$\mathcal{L} = \sum_t \|f_{\theta}(s_{t-1}) - s_t^*\|_2^2 + \kappa \|\theta_{[1:]}\|_1 + \lambda \|\theta_{[1:]}\|_2^2$$



Extended version with more experiments on our project website:

<https://sites.google.com/usc.edu/neuralsim>



github.com/google-research/tiny-differentiable-simulator

