

Benchmarking the Current State-of-the-Art in Reservoir Computing

Bachelor's Thesis

Data-driven predictive modeling has for a long time been a farfetched goal of computational scientists in many fields. Where reduced models attempt to construct faster models by reducing the order (ROMs), reducing the modes (DMD), or simulating on a latent dimension with less resolution (NODE), data-driven predictive modeling attempts to train surrogate models, often neural networks nowadays, to predict the future time-evolution of the physical problem in a data-driven fashion without pre-reducing the order, modes, or similar. [Reservoir Computing](#) is one of the most prominent, and exciting of these data-driven predictive models. It is closely related to recurrent neural networks, which originate from [natural language processing](#), and works by feeding an input signal into the “reservoir”, which is at its core a black-box algorithm with fixed parameters and structure. The output of this black-box algorithm is a latent vector of predefined size, which is then transformed by a readout neural network to the desired output. The core idea here is that if the reservoir is chosen wisely, e.g. containing loosely physics-based updates, the readout neural network can be trained in a data-efficient fashion and much faster than traditional and competing end-to-end neural network approaches. However, most of the current successes of reservoir computing lie in relatively simple physical systems.

In this thesis, we will develop a unified PyTorch-codebase for reservoir computing, based on previous work, and benchmark & probe reservoir computing's performance on Fluid dynamics systems of ever-increasing complexity.

Tasks:

- Begin with the [codebase](#) of [“Next Generation Reservoir Computing”](#) and familiarize yourself with its layout, and data pipeline to apply it ever more difficult fluid dynamics problems.
 - [Burgers](#)
 - [Kuramoto-Sivashinsky](#)
 - [Fluid flow around a rotating cylinder](#)
- Extend the codebase to be able to benchmark different reservoir computing approaches against each other

Requirements:

- Ability to work independently.
- First attempts at playing with PyTorch are beneficial.
- Curiosity to experiment with the approaches, and latent curiosity to dig deeper to understand the strengths and weaknesses of the approach on Fluid Dynamics problems.