

Assignment 3

Forecasting Chaos with Neural Networks

DUE: Friday, March 31, 2023 at 11:59 pm PST
Late assignments will **NOT** be accepted

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The Lorenz equations are possibly the most famous system of ordinary differential equations that exhibit chaotic dynamics. They are given by

$$\begin{aligned}\frac{dx}{dt} &= 10(y - x) \\ \frac{dy}{dt} &= x(\rho - z) - y \\ \frac{dz}{dt} &= xy - \frac{8}{3}z\end{aligned}\tag{1}$$

where $\rho > 0$ is a system parameter. Some background information, figures, and even code snippets for the Lorenz equations are given on the [wikipedia page](#).

Fix a small $\Delta t > 0$. Your task is to train a neural network to take solutions

$$(x(t), y(t), z(t))$$

to

$$(x(t + \Delta t), y(t + \Delta t), z(t + \Delta t))$$

for $\rho = 10, 28$, and 40 . You want to only train one neural network that works for all three parameter values, so your input will be four-dimensional: 3 state variables (x, y, z) and 1 parameter value ρ , with the output being the same. Passing data through the network should not change the parameter value ρ fed into it, but its value will influence how the state variables are advanced. Experiment with the following:

1. The loss function that provides the best predictions.
2. By successively feeding the output of the network back into the network we can advance the predictions Δt time units each iteration. How far into the future do your predictions remain accurate at the trained parameter values $\rho = 10, 28$, and 40 ? What happens if you use a bigger or smaller Δt ?
3. How well does your trained neural network predict the future state for parameter values outside the training data? Try $\rho = 17$ and $\rho = 35$.

After completing the above, fix $\rho = 28$ and train a neural network to identify when a transition from one lobe to another is imminent. Determine how far in advance you can make this prediction. Note: you will have to label the transitions in a test set in order to do this task.

Comment: You may generate your training data using whatever method is easiest to you. That is, using a built-in ODE integrator such as `ode23/ode45` in MATLAB or `odeint` in `scipy` is perfectly acceptable (and probably more accurate). If you are going to simulate the manually, you must at least use a [Runge-Kutta](#) scheme - Euler stepping is not acceptable as it will be either too inaccurate or too slow.