Notes for differential equation predictor neural network

DIFFERENTIAL EQUATIONS

These notes contain the nonlinear differential equations that the neural network in Solving_ODEs_with_NNs.ipynb is trained on. The equations are

$$\frac{d\rho_{ge}}{dt} = (i\Delta - 1) \rho_{ge} - i(2\rho_{ee} - 1) \left[\Omega + (V + i\Gamma)\rho_{ge}\right],$$

$$\frac{d\rho_{ee}}{dt} = -2\rho_{ee} + 2\operatorname{Im}[\Omega^*\rho_{ge}] - 4\Gamma|\rho_{ge}|^2.$$
(1a)

$$\frac{d\rho_{ee}}{dt} = -2\rho_{ee} + 2\operatorname{Im}[\Omega^*\rho_{ge}] - 4\Gamma|\rho_{ge}|^2.$$
(1b)

where ρ_{ge} is a complex parameter that can be split into a real and imaginary part. The parameters Ω and Δ are variables that are changed when training the neural network. The parameters V and Γ have fixed values V=20.4and $\Gamma = 22.9$. Eqs. (1) can be used to describe certain physical systems, such as a cloud of atoms with two-energy levels (e.g., the atom can be excited or unexcited), which are then driven by a laser with strength Ω and frequency Δ . The coefficients V and Γ are determined by the sum of interactions between the atoms.

At large enough times, Eqs. (1) eventually reach the steady-state, i.e., $d\rho_{qe}/dt = 0$ and $d\rho_{ee}/dt = 0$. For certain values of Ω and Δ , there can be two steady-states, which is known as bistability. Because Eqs. (1) are fairly simple, we can actually determine the steady-state solutions semi-analytically, and find that

$$\rho_{ge} = \frac{i\Omega(2\rho_{ee} - 1)}{i\left[\Delta - (2\rho_{ee} - 1)V\right] - \left[1 - (2\rho_{ee} - 1)\Gamma\right]},\tag{2}$$

while $(2\rho_{ee}-1)$ obeys the cubic equation

$$[\Gamma^{2} + V^{2}](2\rho_{ee} - 1)^{3} + [\Gamma^{2} + V^{2} - 2\Delta V - 2\Gamma](2\rho_{ee} - 1)^{2} + [\Delta^{2} + 1 + 2|\Omega|^{2} - 2\Delta V - 2\Gamma](2\rho_{ee} - 1) + (\Delta^{2} + 1) = 0.$$
(3)

Eq. (3) has two real positive solutions when the differential equations display bistability.