PINNs: - Data-driven discovery of p.d.e.s (system identification) - Data-driven solution of -11-

e.g. Burger's equation:
$$N(u; \eta) = \lambda, u \frac{\partial u}{\partial x} + \lambda_2 \frac{\partial^2 u}{\partial x^2}, \lambda = (\lambda_1, \lambda_2)$$

Given noisy measurements of ulpi)] compute:

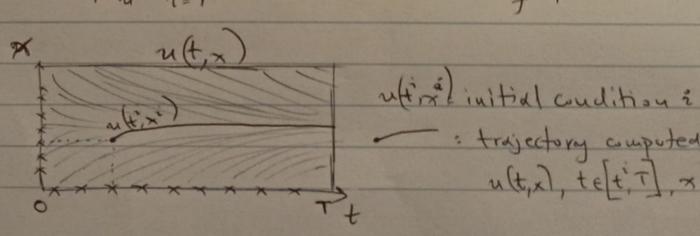
$$(\lambda)$$
 $u(t,x)$, $x \in \Omega$, $t \in [0,T]$

Let f:= du + N(u). N is known. If we approximate uf,x) by an NN, the f is called a "physics-informed NN".

e.g. Duger's equation:
$$f:=dn+u\partial_{\pi}-(0.01/\pi)\cdot\frac{\partial^2u}{\partial x^2}$$

Let MSE be the training loss function.

 $MSE := MSE_{u} + MSE_{f}$ $= \frac{1}{N_{u}} \frac{N_{u}}{1=1} \left[\frac{1}{N_{u}} \left(\frac{1}{t_{u}}, x_{u}^{i} \right) - u^{i} \right]_{x}^{2} + \frac{1}{N_{f}} \frac{1}{1=1} \left[\frac{1}{f} \left(\frac{1}{t_{f}}, x_{f}^{i} \right) \right]_{x}^{2}$



trajectory computed

u(t,x), te[t',T], xess