

## Conservation laws

$$\underline{u}_t + \nabla \cdot F(\underline{u}) = 0$$

$\underline{u}$  = conserved quantity

### Example

let  $q(x,t)$  = concentration

then  $\int_{x_1}^{x_2} q(x,t) dx$  = total mass in  $[x_1, x_2]$  at "t"

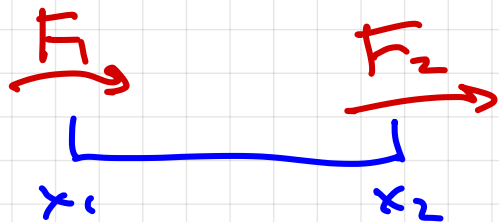
→ mass only changes due to flux at the ends

$$\frac{d}{dt} \int_{x_1}^{x_2} q(x,t) dx = F_1(t) - F_2(t)$$

Fluid?  $F = u(x,t) \cdot q(x,t)$

velocity  $\frac{m}{s}$       density  $\frac{kg}{m}$

$$= \underline{u} q(x,t) \text{ for constant velocity}$$



$$\begin{aligned}
 \rightarrow \frac{d}{dt} \int_{x_1}^{x_2} q(x,t) dx &= f(q(x_1,t)) - f(q(x_2,t)) \\
 &= -f(q(x,t)) \Big|_{x_1}^{x_2} \\
 &= - \int_{x_1}^{x_2} \frac{\partial f(q(x,t))}{\partial x} dx
 \end{aligned}$$

$$\rightarrow \int_{x_1}^{x_2} \frac{\partial q(x,t)}{\partial t} + \frac{\partial f(q(x,t))}{\partial x} dx = 0$$

---


$$q_t + (f(q))_x = 0$$

$$u_t + \underset{\substack{\uparrow \\ \partial_x}}{\nabla \cdot} \left( \frac{u^2}{2} \right) = 0$$


---

$$u_t + \nabla \cdot (au) = 0$$


---

$$\begin{aligned} \int (u_t + \nabla \cdot F)v &= 0 \\ &= \int u_t v + \int_{\partial \Omega} n \cdot F v - \int_{\Omega} F \cdot \nabla v = 0 \end{aligned}$$

# PINN

main challenges:

1. nothing PDE specific  
→ how to enforce conservation, e.g.?
2. cost  
→ everything is monolithic

Conservative physics-informed neural networks on discrete domains for conservation laws: Applications to forward and inverse problems

Ameya D. Jagtap<sup>a</sup>, Ehsan Kharazmi<sup>a</sup>, George Em Karniadakis<sup>a,b,\*</sup>

2020

→ cPINN

↓  
physics constrained NN

Extended physics-informed neural networks (XPINNs) : A generalized space-time domain decomposition based deep learning framework for nonlinear partial differential equations

Ameya D. Jagtap<sup>\* 1</sup>  
George Em Karniadakis<sup>1</sup>

2021

→ generalizes cPINN  
XPINN

**D3M: A Deep Domain Decomposition Method for Partial Differential Equations**

2019

→ D3M

KE LI<sup>1,2,3</sup>, (Student Member, IEEE), KEJUN TANG<sup>1,2,3</sup>, TIANFAN WU<sup>4</sup>, AND QIFENG LIAO<sup>1</sup>

Parallel Physics-Informed Neural Networks via Domain Decomposition

Khemraj Shukla, Ameya D. Jagtap, George Em Karniadakis\*

2021?

## cPINN

- use domain decomp to enforce locality
- provide information on fluxes at interfaces
- per domain NN selection
- adaptive activations

