

Neural Network-based Detection of Taylor Vortices in Annular Flow Systems

Exposé for Master Thesis - Initial Presentation

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Turbulent Flows and Vortices

- Vortical formations are crucial components in the dynamics of turbulence, contributing to processes such as the generation of turbulence, amplification of mass, heat, and momentum transport.
- A vortex is characterized by swiriling motion of fluid around a central region.
- Translating this feature into a formal definition is challenging.

"A vortex exists when instantaneous streamlines mapped onto a plane normal to the vortex core exhibit a roughly circular or spiral pattern, when viewed from a reference frame moving with the center of the vortex core."



Taylor-Couette flow

- a Fluid dynamic phenomenon that occurs when a fluid is passing between two coaxial-rotating cylinders.
- Inner cylinder is typically rotating faster than outer cylinder.
- Dimensionless control parameters like Re, ω_{inner} and ω_{outer} are key factors

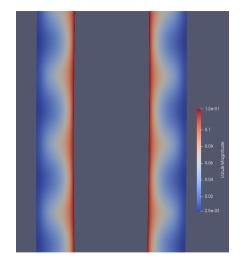


Abbildung: Taylor-couette flow at time-step 19



Vortex Detection Algorithems

- Vortex identification methods are categorized into 3 taxonomies:
 - Region/Line
 - Eulerian/Lagrangian.
 - Local/Global.
- One of widly used algorithms is λ_2 method:
 - $-\nabla \overline{u}$ is decomposed into symmetric(S) and $anti-symmetric(\Omega)$ parts.
 - three eigenvalues of $S^2+\Omega^2$ to be calculated.
 - A point int the velocity field is a part of vortex core if at least 2 of them are negative i.e. $\lambda_2 < 0$



Objectives

Governing equ:

$$\frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla)\mathbf{u} = -\frac{1}{\rho}\nabla p + \nu \nabla^2 \mathbf{u} + \mathbf{f}$$
 (1)

w.r.t boundary conditions:

$$\mathbf{u} = \mathbf{u}_0$$
 at Γ_1 (2)

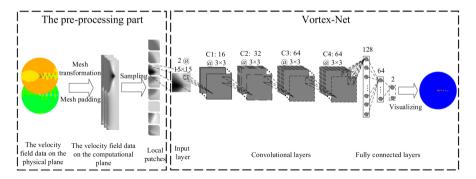
$$\mathbf{u} = 0$$
 at Γ_2 (3)

$$\frac{\partial p}{\partial n} = 0 \quad \text{at} \quad \Gamma_3 \tag{4}$$



Litrature Review

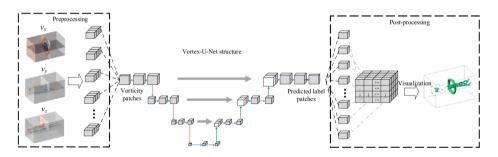
(a) (Liang et al., 2018), a CNN-based vortex identification method to use both local and global information of flow field.





Litrature Review

(b) (Deng et al., 2022), replacing the fully-connected NN with a segmented network to reduce the computational complexity.



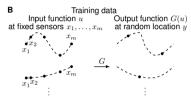


Approach

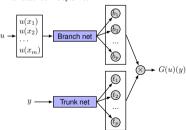
- (a) Data preparation:
 - Running the Taylor merve case for different values of $\{\omega_{\it inner}, \omega_{\it outer}, Re\}$
 - Collecting the velocity fields and λ_2 vectors for all time steps.
- (b) Model development:
 - Trying to Learn the mapping from a function in field to λ_2 method for detecting vortices.



Deep operator networks(Lu, Jin, Pang, Zhang & Karniadakis, 2021) (DeepONets)



D Unstacked DeepONet

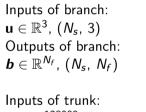


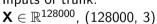
- Each input function u is evaluated at fixed sensor points $\{x_1, x_2, \dots, x_m\}$
- y with d components and u(x_i) for i = 1, 2, ..., m are not matched. Therefor, it is needed to use two subnets. Branch for encoding input function at sensor points - Trunk for the location to evaluate output function

$$G(u)(y) \approx \sum_{k=1}^{p} b_k t_k + b_0 \qquad (5)$$



Deep operator networks (DeepONets): prediction of λ_2

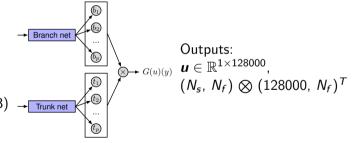




Outputs of trunk:

Outputs of trunk

$$\boldsymbol{t} \in \mathbb{R}^{N_f}$$
, (128000, N_f)





Motivation

- Learning the λ_2 operator for detecting vortices within a field.
- Replacing the Finit-Element mothods with NN-surrogate model
- Elinimate the need for manual detection of vortices in flow system using visualization



Timeline

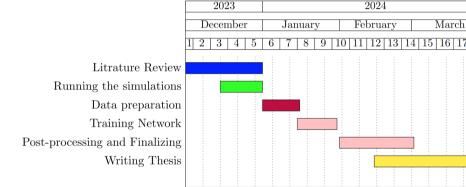


Abbildung: Timeplan



Thank you!

Any questions?

TU Clausthal

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