



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

Exposé for Master Thesis - Initial Presentation

Mahyar Alikhani, Institute of Applied Mechanics

12. Dezember 2023



Contents

Problem Statement

Taylor-Couette flow

Vortex Detection Algorithms

Task definition and objective

Literature Review

Approache

Deep Operator Learning (DeepOnet)

Motivation

Time Planning

Literatur15

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 swirling 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

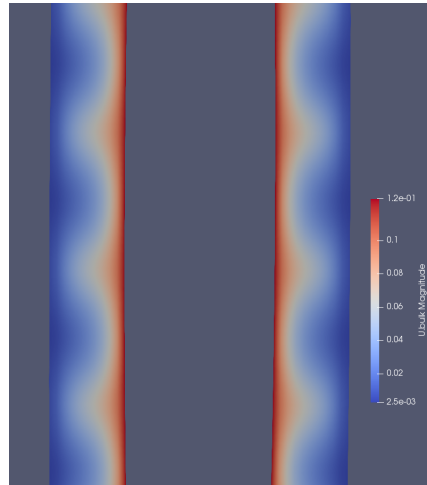


Abbildung: Taylor-couette flow at time-step 19

Vortex Detection Algorithms

- Vortex identificaion methods are categorized into 3 taxonomies:
 - **Region/Line**
 - **Eulerian/Lagrangian.**
 - **Local/Global.**
- One of widly used algorithms is λ_2 method:
 - $\nabla \bar{u}$ is decomposed into *symmetric*(S) and *anti-symmetric*(Ω) 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:

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

- w.r.t boundary conditions:

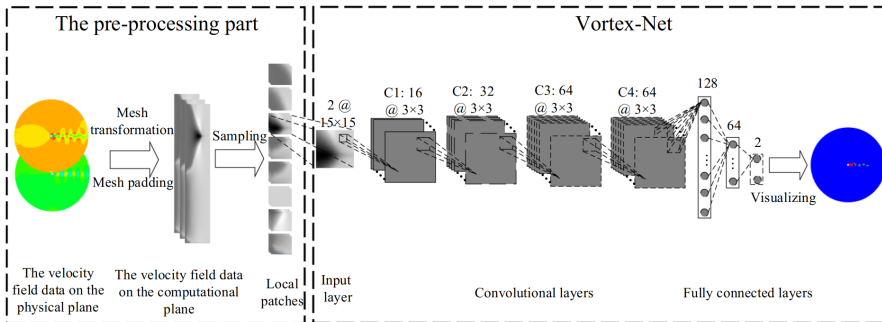
$$\mathbf{u} = \mathbf{u}_0 \quad \text{at} \quad \Gamma_1 \quad (2)$$

$$\mathbf{u} = 0 \quad \text{at} \quad \Gamma_2 \quad (3)$$

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

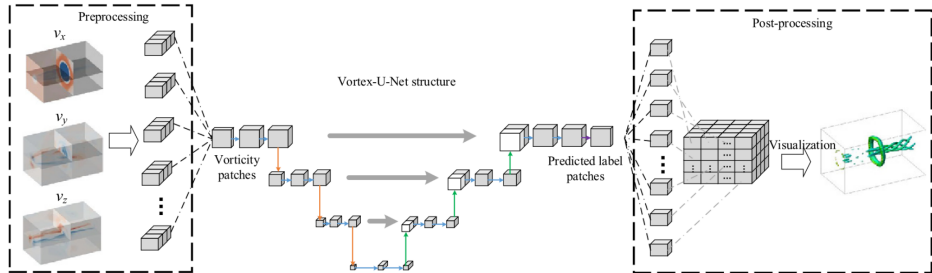
Literture Review

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



Literature Review

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



Approach

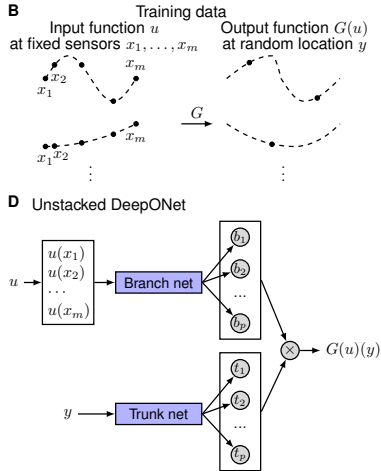
(a) Data preparation:

- Running the *Taylor merve* case for different values of $\{\omega_{inner}, \omega_{outer}, Re\}$
- 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)



- 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 $i = 1, 2, \dots, m$ are not matched. Therefore, it is needed to use two subnets. **Branch** for encoding input function at sensor points - **Trunk** for the location to evaluate output function

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

Deep operator networks (DeepONets): prediction of λ_2

Inputs of branch:

$$\mathbf{u} \in \mathbb{R}^3, (N_s, 3)$$

Outputs of branch:

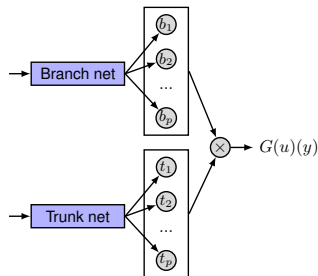
$$\mathbf{b} \in \mathbb{R}^{N_f}, (N_s, N_f)$$

Inputs of trunk:

$$\mathbf{X} \in \mathbb{R}^{128000}, (128000, 3)$$

Outputs of trunk:

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



Outputs:

$$\mathbf{u} \in \mathbb{R}^{1 \times 128000},$$

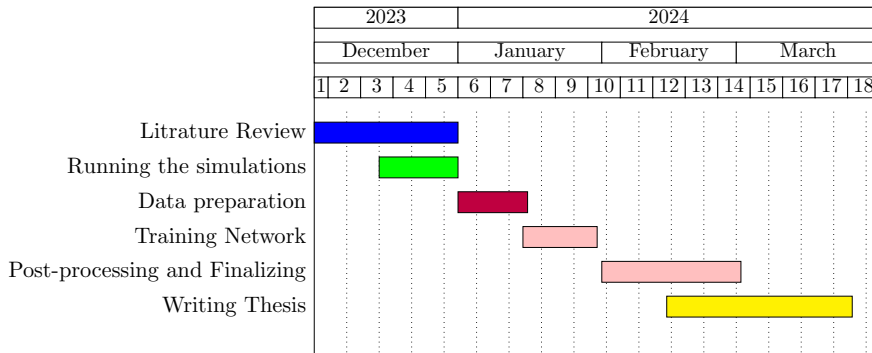
$$(N_s, N_f) \otimes (128000, N_f)^T$$



Motivation

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

Timeline





Thank you!

Any questions?

- Deng, L., Bao, W., Wang, Y., Yang, Z., Zhao, D., Wang, F., ... Guo, Y. (2022). Vortex-u-net: An efficient and effective vortex detection approach based on u-net structure. *Applied Soft Computing*, 115, 108229. Zugriff auf <https://www.sciencedirect.com/science/article/pii/S1568494621010620> doi: <https://doi.org/10.1016/j.asoc.2021.108229>
- Liang, D., Wang, Y., Liu, Y., Wang, F., Li, S. & Liu, J. (2018, 10). A cnn-based vortex identification method. *Journal of Visualization*, 22. doi: 10.1007/s12650-018-0523-1
- Lu, L., Jin, P., Pang, G., Zhang, Z. & Karniadakis, G. E. (2021). Learning nonlinear operators via DeepONet based on the universal approximation theorem of operators. *Nature Machine Intelligence*, 3 (3), 218–229.