



# 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 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  $Re$ ,  $\omega_{inner}$  and  $\omega_{outer}$  are key factors

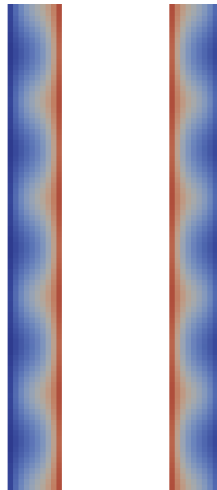


Abbildung: Taylor-couette flow at time-step 19

## Vortex Detection Algorithms

- Vortex identification methods are categorized into 3 taxonomies:
  - **Region/Line**
  - **Eulerian/Lagrangian.**
  - **Local/Global.**
- One of widely used algorithms is  $\lambda_2$  method:
  - $\nabla \bar{u}$  is decomposed into *symmetric*( $S$ ) and *anti-symmetric*( $\Omega$ ) parts.
  - three eigenvalues of  $S^2 + \Omega^2$  to be calculated.
  - A point in 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:

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

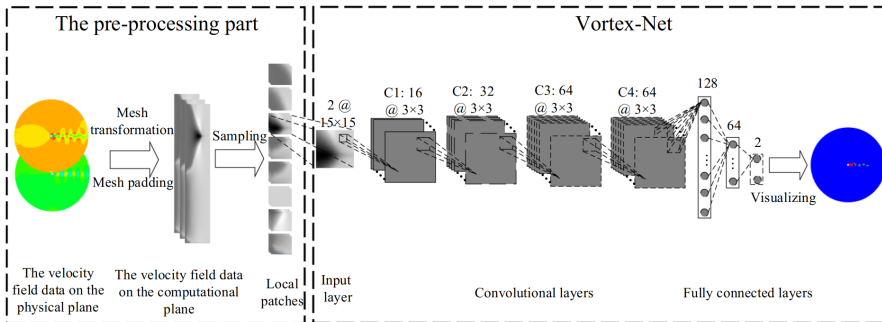
- w.r.t boundary condition:

$$\mathbf{u}(\mathbf{x} + \mathbf{L}) = \mathbf{u}(\mathbf{x}), \quad (2)$$

where  $\mathbf{L}$  is the periodicity vector

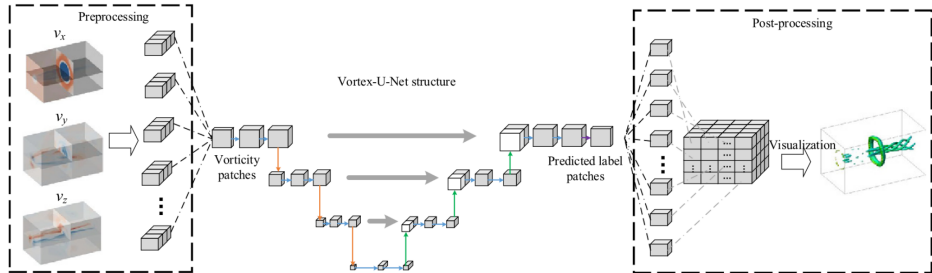
## Literture Review

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



## Literture Review

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





## Approach

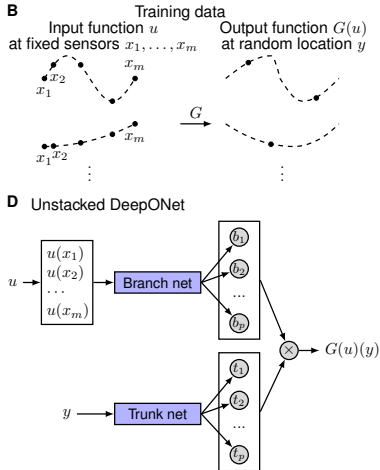
### (a) Data preparation:

- Running the *Taylor-couette* case for different values of  $\{\omega_{inner}, \omega_{outer}, Re\}$
- Collecting the velocity fields and  $\lambda_2$  vectors for all time steps.
- Data augmentation techniques (Berenjkoub, Chen & Günther, 2020) like *Linear Transformation* can be used, to compensate the lack of training data.

### (b) Model development:

- Trying to Learn the mapping from a function in field to  $\lambda_2$  method for detecting vortices.
- The objective of trained network is to perform a classification based on the predicted  $\lambda_2$  values.

## 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

$$G(u)(y) \approx \sum_{k=1}^p b_k t_k + b_0 \quad (3)$$

## Deep operator networks (DeepONets): prediction of $\lambda_2$

Inputs of branch:

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

Outputs of branch:

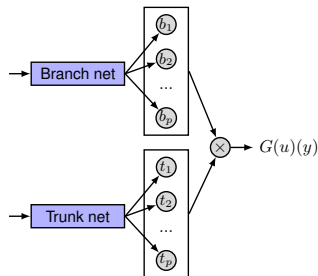
$$\mathbf{b} \in \mathbb{R}^{N_f}, (N_s, N_f)$$

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:

$$\lambda_2 \in \mathbb{R}^{1 \times 128000},$$

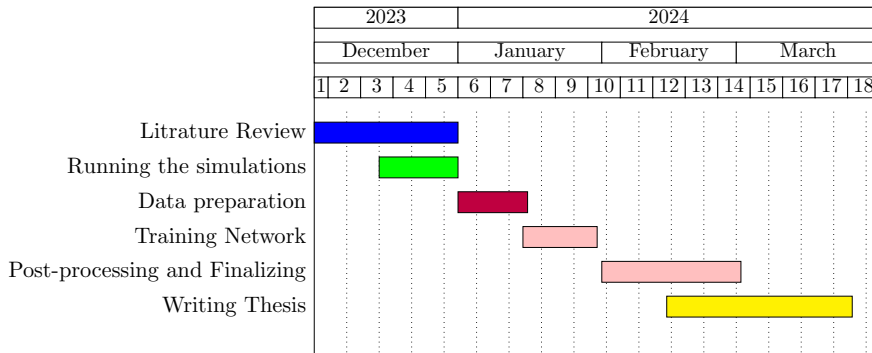
$$(N_s, N_f) \otimes (128000, N_f)^T$$



## Motivation

- Learning the  $\lambda_2$  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?

- Berenjkoub, M., Chen, G. & Günther, T. (2020, 10). Vortex boundary identification using convolutional neural network.. doi: 10.1109/VIS47514.2020.00059
- 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>
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