

# ModelFLOWS APP

## MODAL DECOMPOSITION

Pattern detection

Reconstruction

Prediction

HOSVD

Data Repairing

HODMD

HODMD

Superresolution

## DEEP LEARNING

Pattern detection

Reconstruction

Prediction

Autoencoders

Superresolution

Full DL

Hybrid



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**HODMD**

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# Motivation

HODMD

Physics of Fluids

LETTER

scitation.org/journal/phf

## On the generation and destruction mechanisms of arch vortices in urban fluid flows

Cite as: Phys. Fluids **34**, 051702 (2022); doi: [10.1063/5.0088305](https://doi.org/10.1063/5.0088305)

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and Ricardo Vinuesa<sup>3,a)</sup> 

### AFFILIATIONS

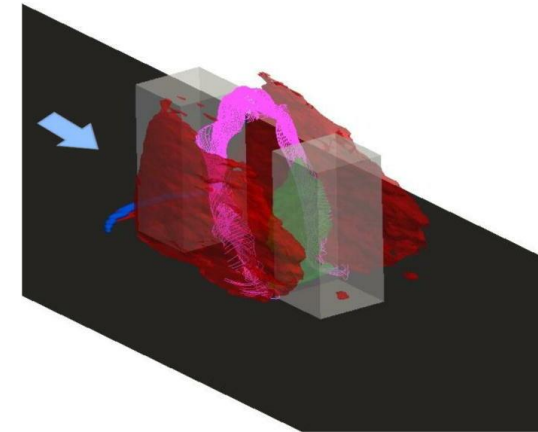
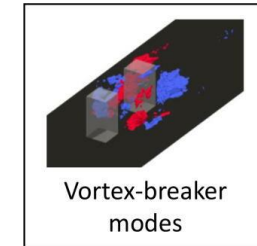
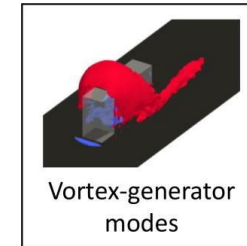
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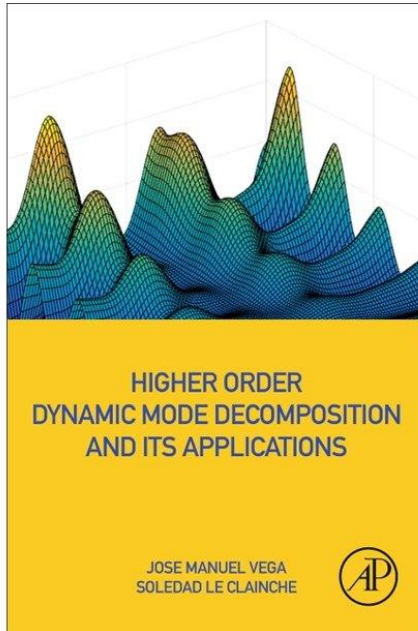
<https://doi.org/10.1063/5.0088305>



Pattern analysis in turbulent complex flows in urban environments.

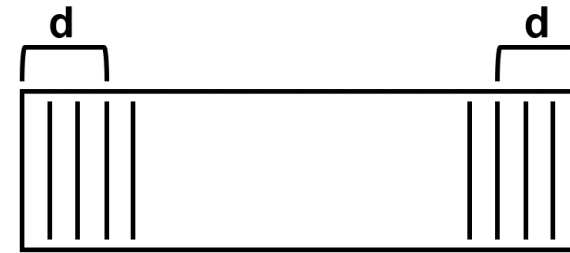
# Methodology

HODMD



Vega, J. M., & Le Clainche, S. (2020). *Higher order dynamic mode decomposition and its applications*. Academic Press.

$$\mathbf{V}_{d+1}^K \cong R_1 \mathbf{V}_1^{K-d} + R_2 \mathbf{V}_2^{K-(d-1)} + \dots + R_d \mathbf{V}_d^{K-1}$$



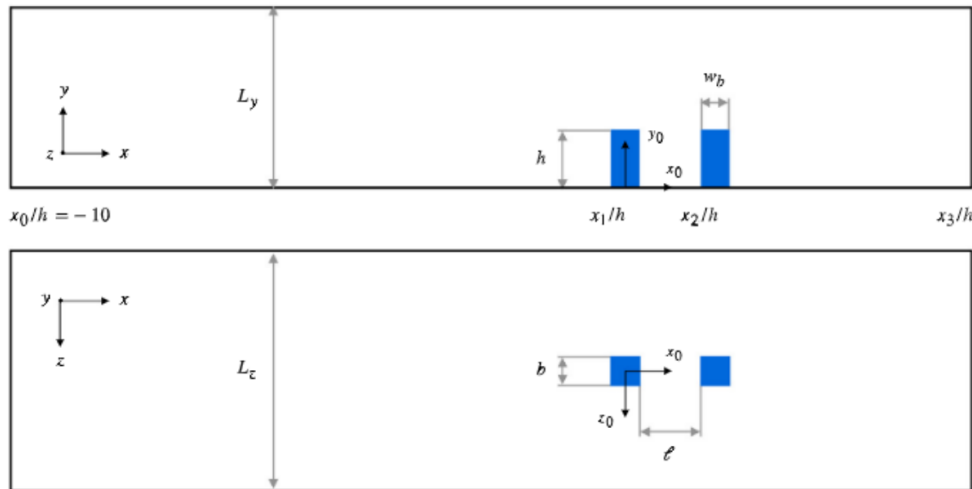
**DMD modes**

$$\mathbf{M}(t) \simeq \sum_{m=1}^M \underbrace{a_m}_{\text{Amplitude}} \underbrace{\mathbf{u}_m}_{\text{Growth-rate}} \underbrace{e^{(\delta_m + i\omega_m)t}}_{\text{Frequency}} \text{ for } t_1 \leq t \leq t_1 + T$$



# Database & Data preparation

HODMD



Spatial dimensions

Snapshots Tensor =  $\{N_v, N_x, N_y, N_z, N_t\}$

Variables

Temporal dimensions

# Database & Data preparation

HODMD

$$\text{Snapshots Tensor} = \{N_v, N_x, N_y, N_z, N_t\} \left\{ \begin{array}{l} - N_v = 3 \\ - N_x = 100 \\ - N_y = 125 \\ - N_z = 50 \\ - N_t = 224 \end{array} \right. \Rightarrow \begin{array}{l} \text{Spatial dimension} = 1875000 \\ \text{Temporal dimension} = 224 \end{array}$$

# Calibration

HODMD

Spatial dimension = 1875000

Temporal dimension = 224

**Tolerance SVD:** How many singular values we retain. In turbulence, different scales.

Values: 1e-2, 1e-3, 1e-4

**Tolerance DMD:** What minimum amplitude is associated to the modes. Different scales.

Values: 1e-2, 1e-3, 1e-4

**Number of windows:** Needed for the HODMD windowing process ( $\sim 10 - 50\% N_t$ ).

Values: 10, 20, 50, 100



ModelFLOWS



# Calibration

HODMD

## *Tolerance SVD:*

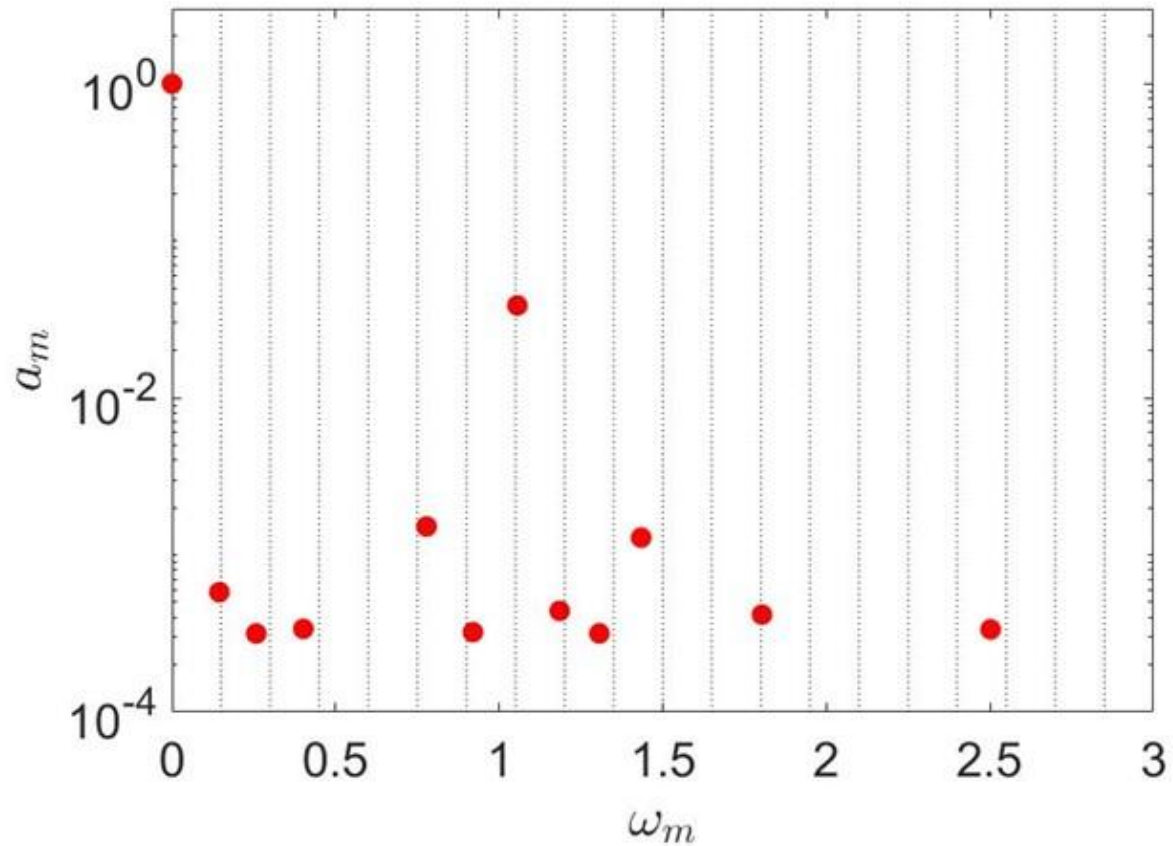
Values: 1e-3

## *Tolerance DMD:*

Values: 1e-3

## *Window size:*

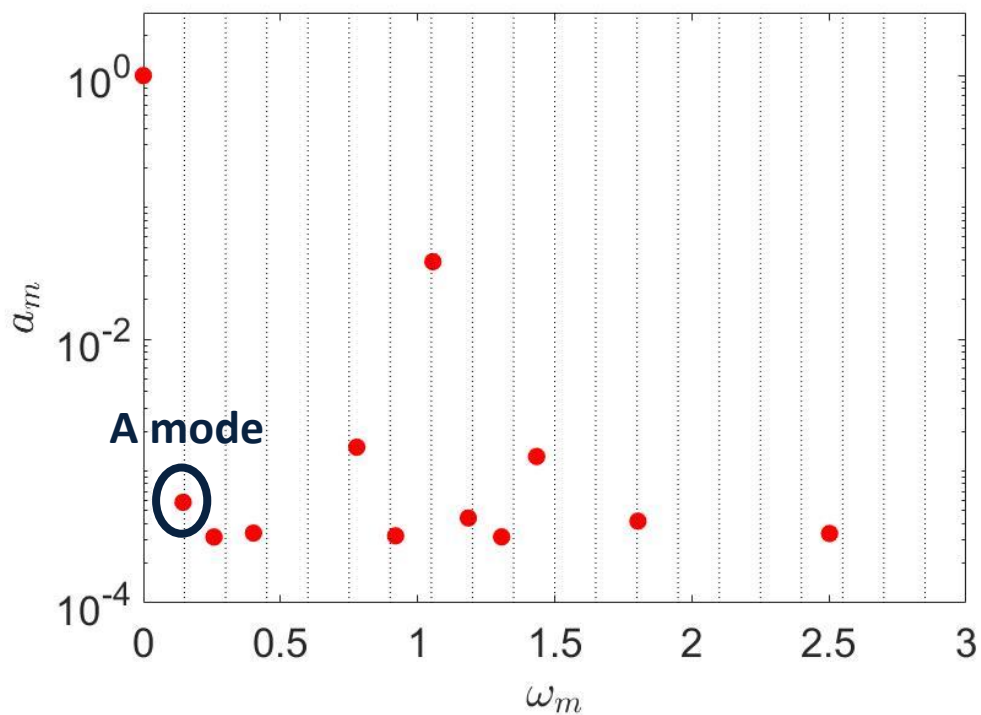
Values: 50



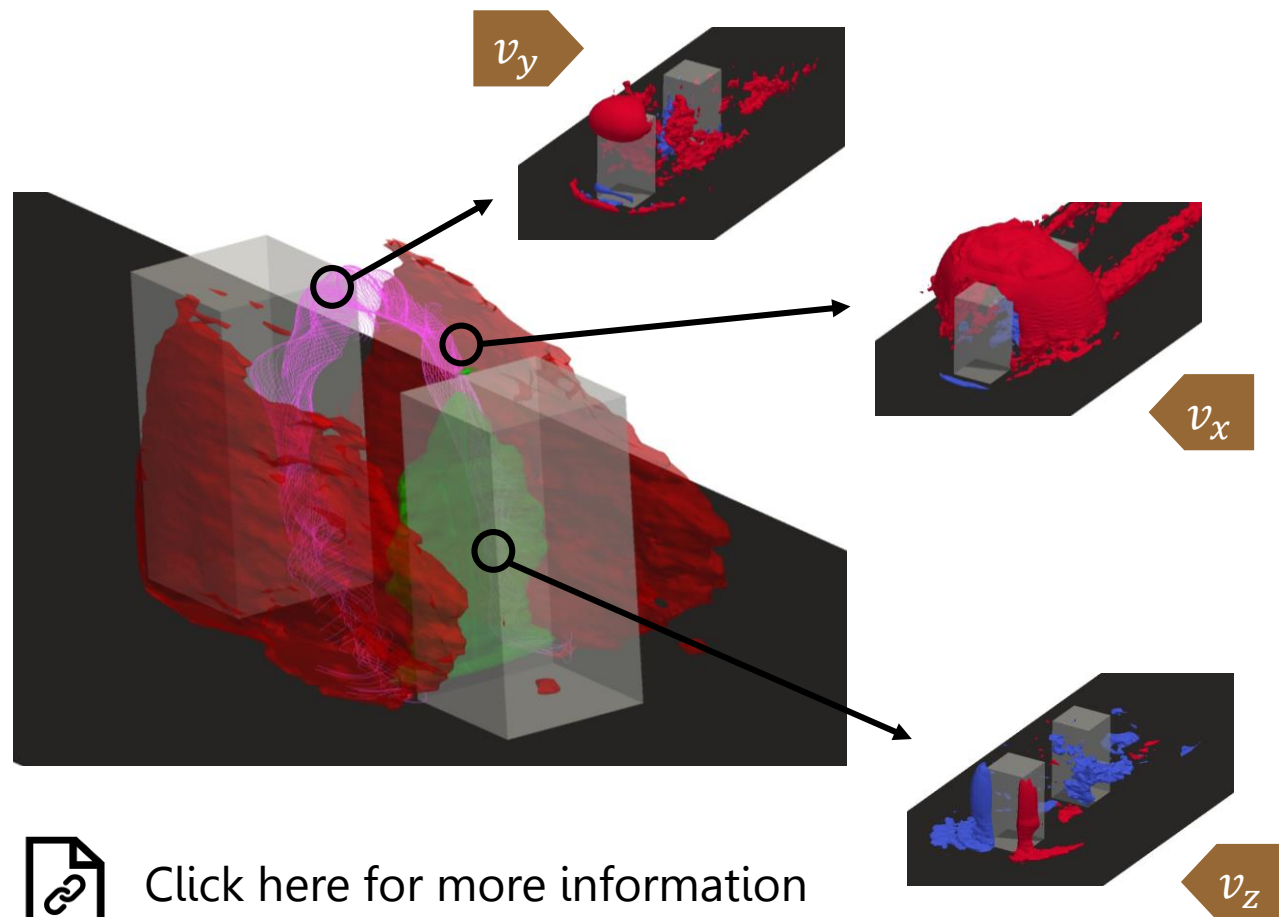


# Results

HODMD

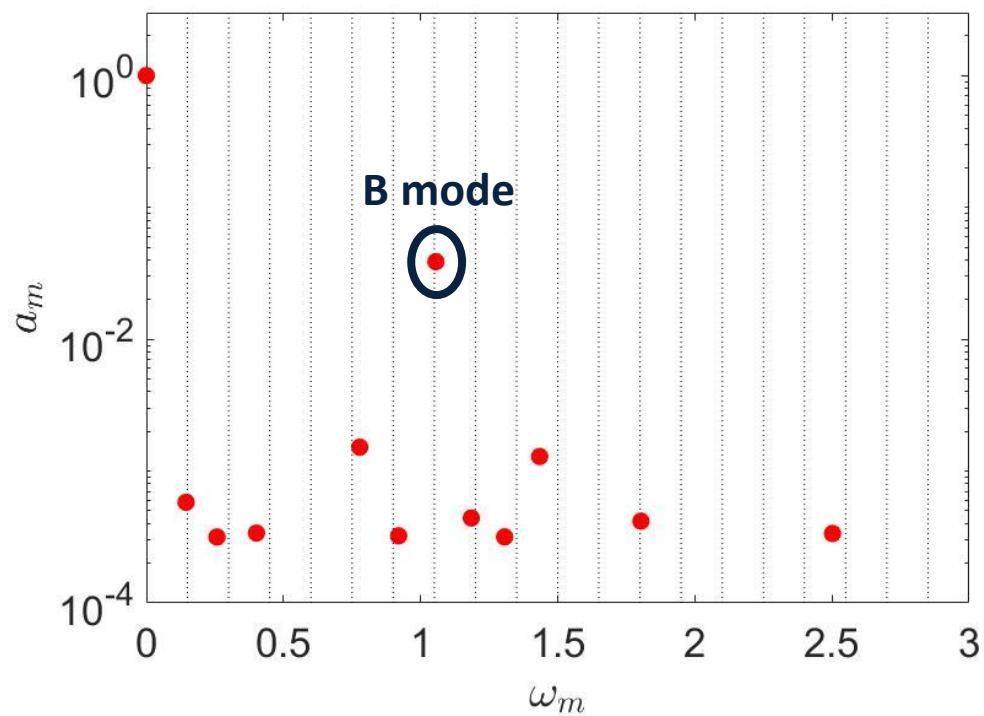


Generator modes



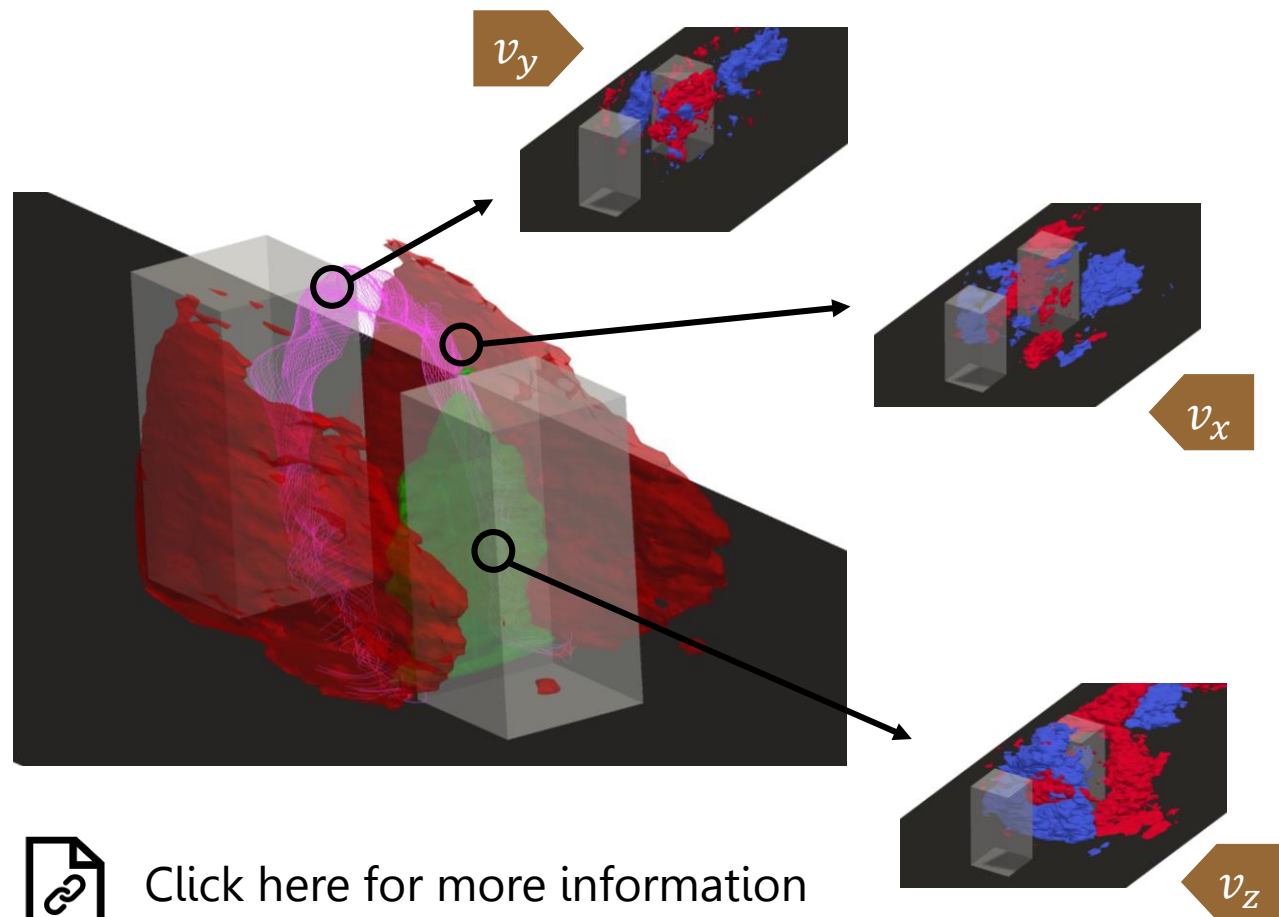
Click here for more information

# Results



Breaker modes

HODMD





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