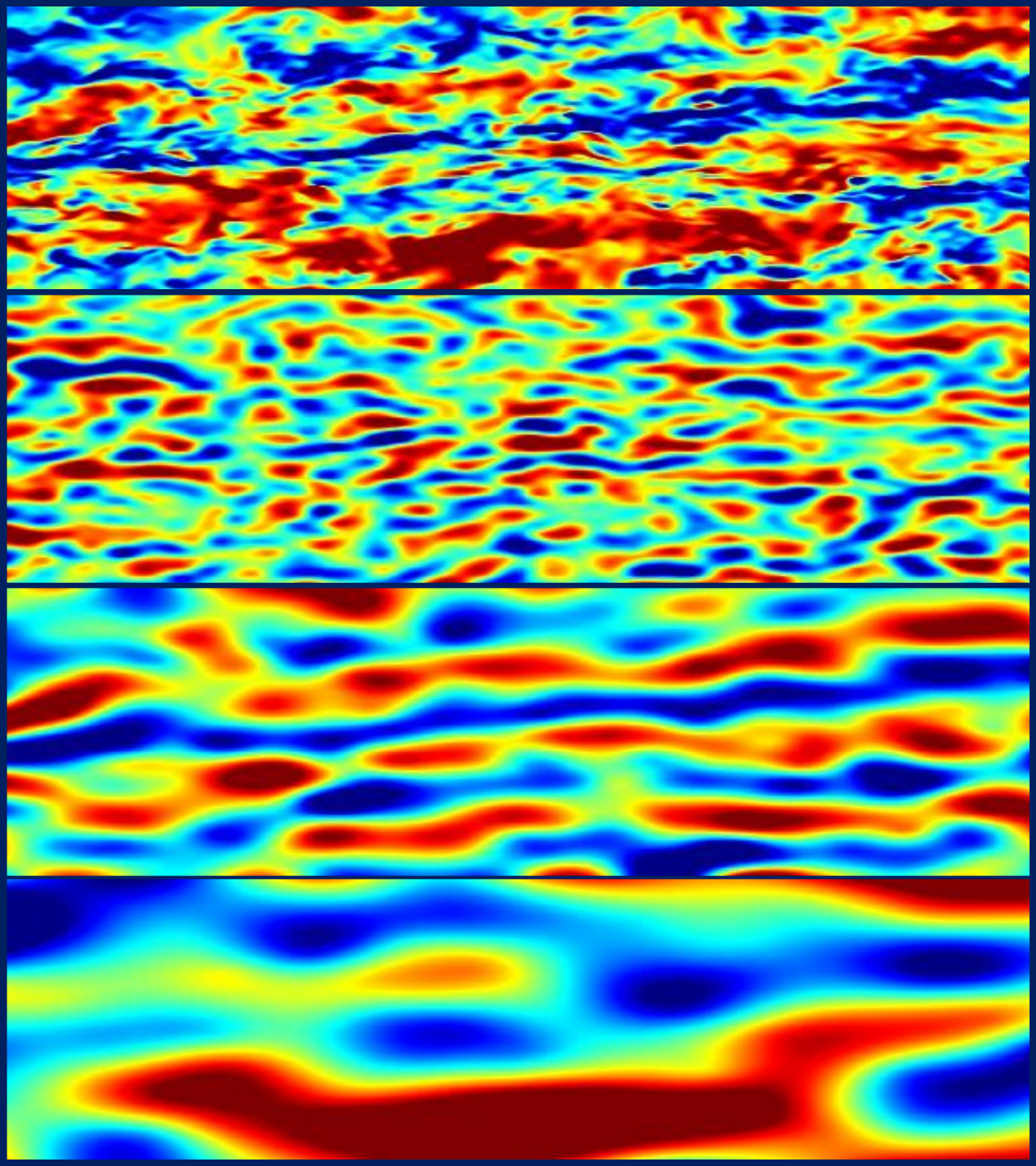


Modal Scale Decomposition in Turbulent Flows

Chair of Fluid Mechanics and Institute of
Aerodynamics
RWTH Aachen University, Germany

Esther Lagemann
née Mäteling



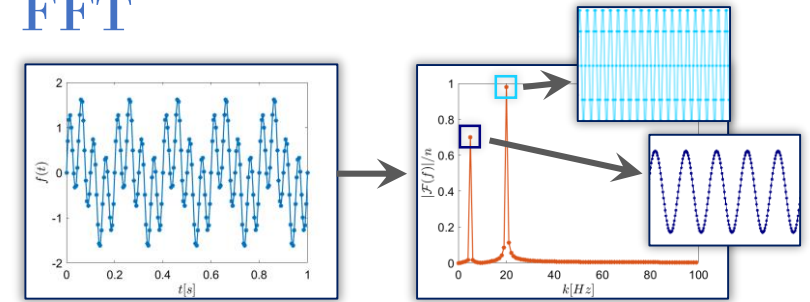
objective:

separate data into **purposeful components**
to gain new insight into **inherent features**

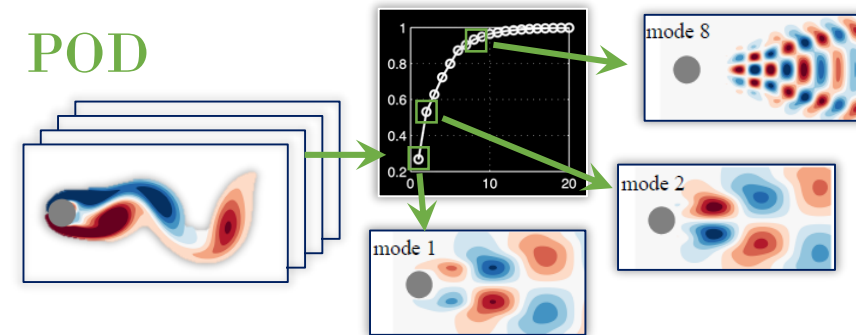
variety of established modal decomposition methods:

- Fast Fourier Transform (**FFT**)
- Singular Value Decomposition (SVD) / Proper Orthogonal Decomposition (**POD**)
- Dynamic Mode Decomposition (**DMD**)
- ... a variety of extensions

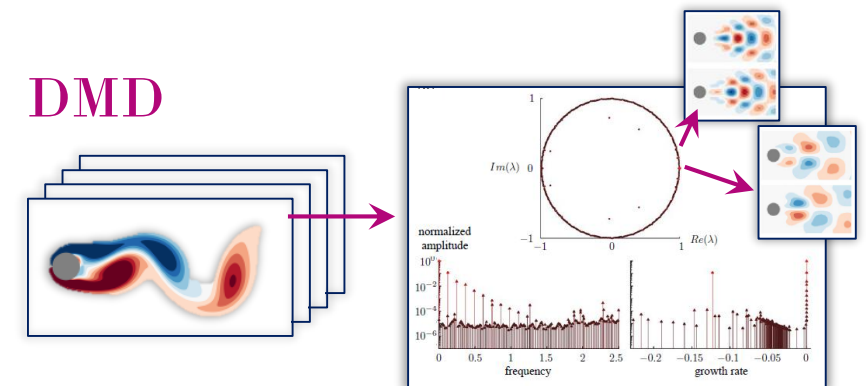
FFT



POD



DMD



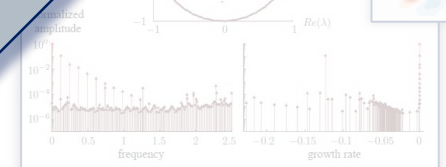
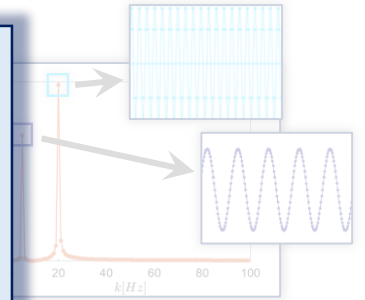
Why do we need alternative methods for turbulent flows?

- unsteady dynamics
- non-linear phenomena
- broadband data
- multi-scale interactions



majority of established decomposition methods fails to capture such characteristics

FFT





EMPIRICAL MODE DECOMPOSITION

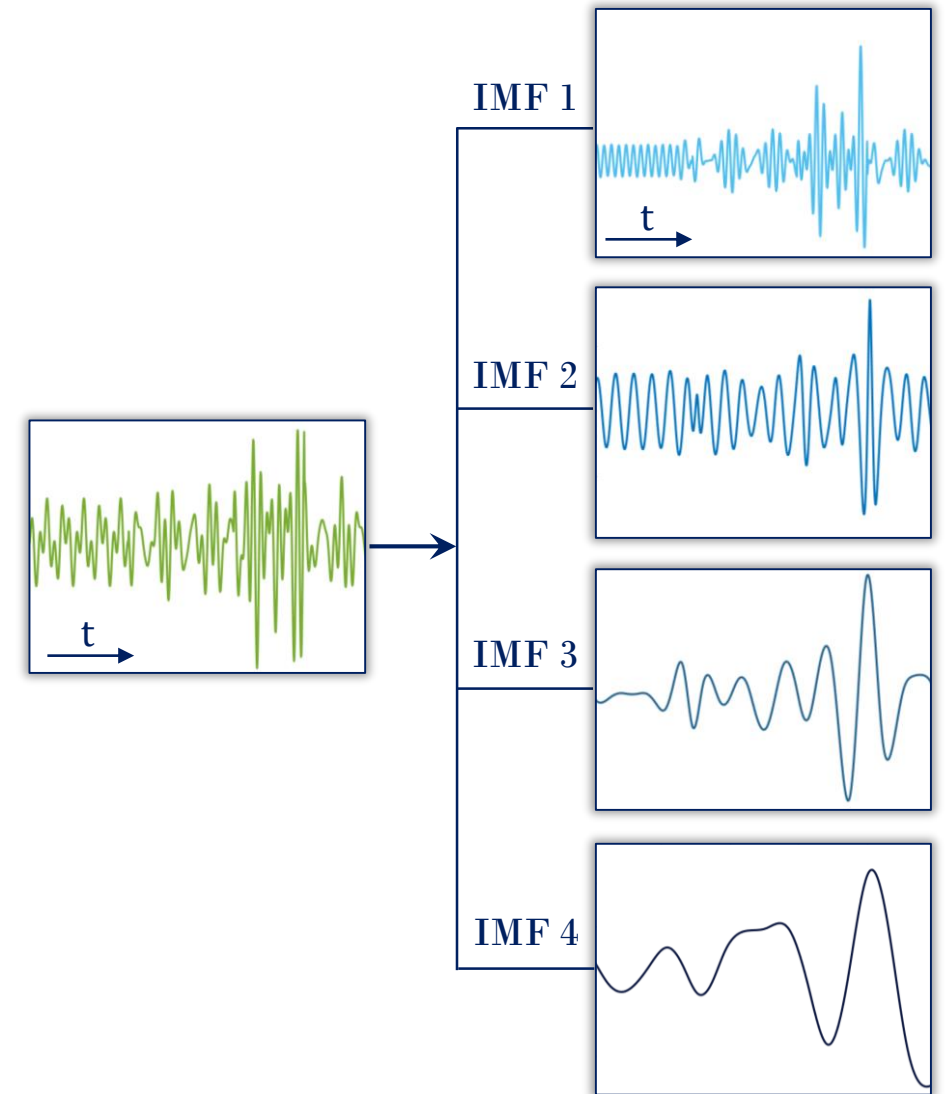
- 1D univariate EMD introduced in 1998
- data-driven decomposition based on intrinsic scales
- modes (IMFs) are sorted with respect to scale size

- ability to process non-linear and unsteady data
- basis system dictated by data
 - modes adaptively biased towards locally dominant frequencies/scales



shortcomings for fluid dynamics

- improper mode alignment
- aliasing in time-frequency domain
- problems of uniqueness
- restricted complexity/dimensionality

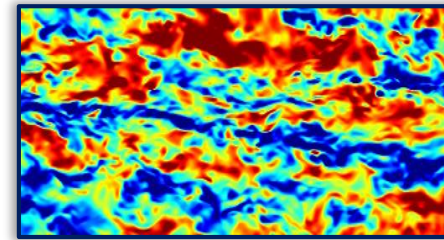




**2D NOISE-ASSISTED
MULTIVARIATE EMPIRICAL
MODE DECOMPOSITION**

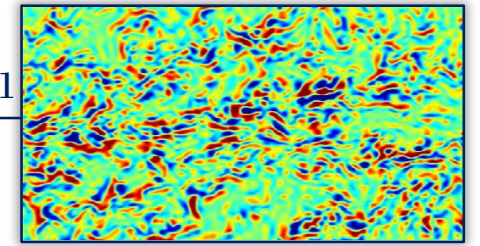
2D Noise-Assisted Multivariate Empirical Mode Decomposition

process 2D snapshots

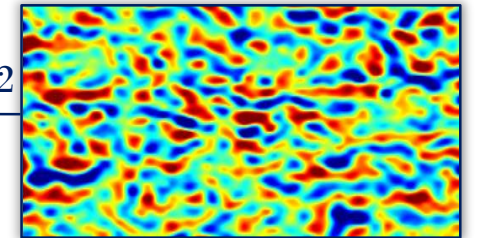


$$u' = f(x, z)$$

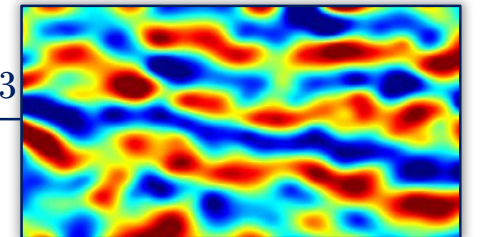
IMF 1



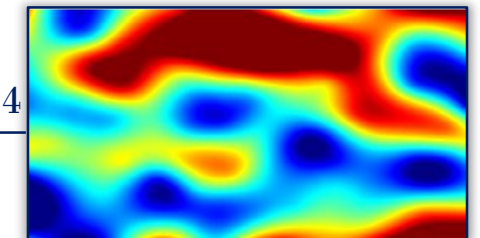
IMF 2



IMF 3



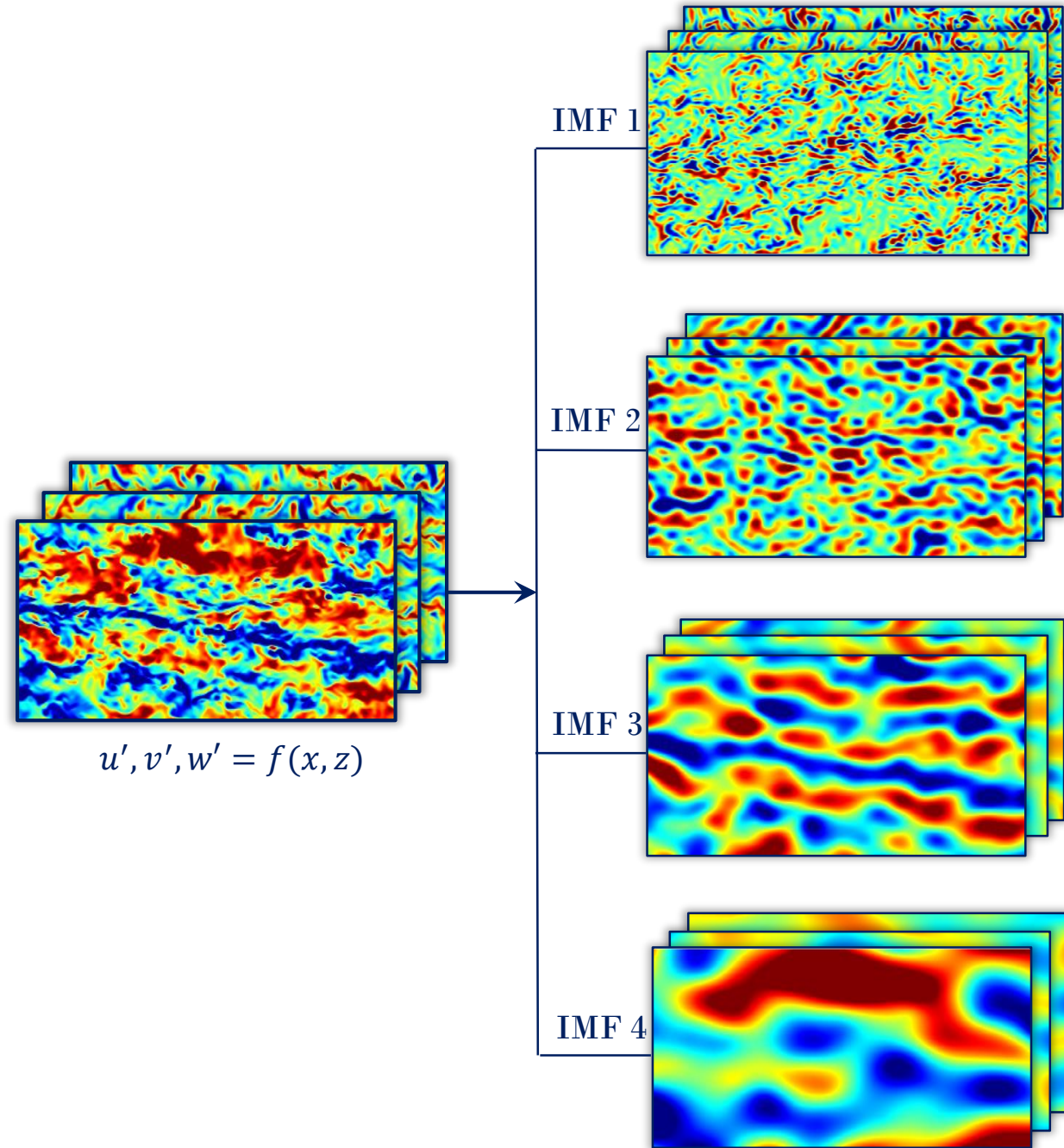
IMF 4



2D Noise-Assisted Multivariate Empirical Mode Decomposition

process 2D snapshots

find & align common features
within multivariate data, e.g.
several velocity components

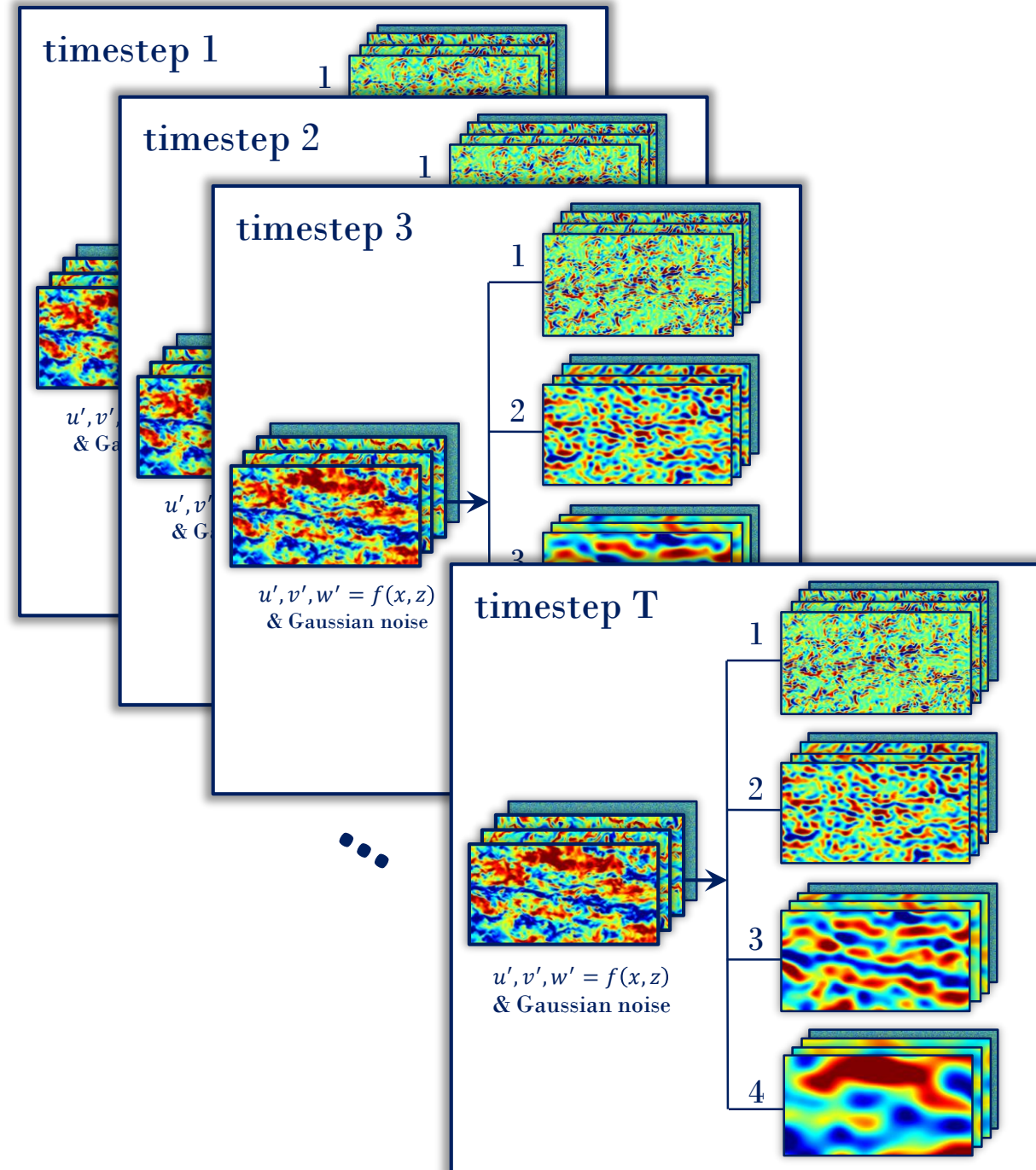


2D Noise-Assisted Multivariate Empirical Mode Decomposition

process 2D snapshots

find & align common features
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several velocity components

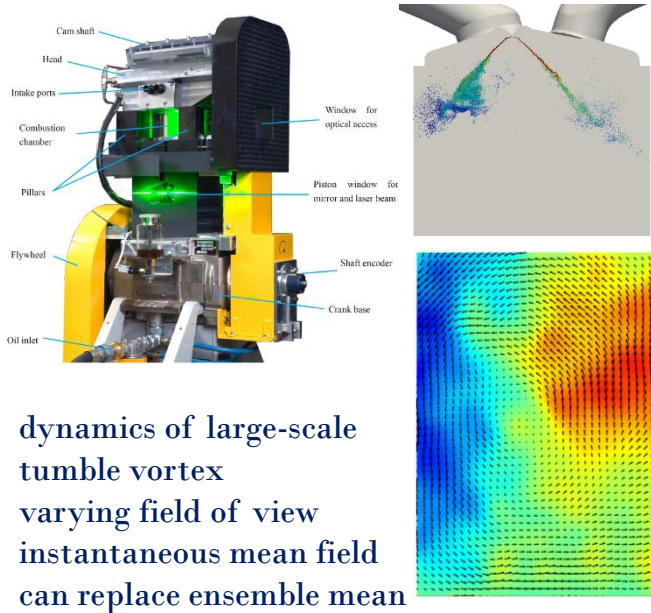
improve mode separation and
coherence across decompositions
by variates of Gaussian noise





**2D NA-MEMD
FOR TURBULENT FLOWS**

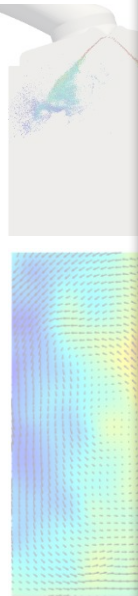
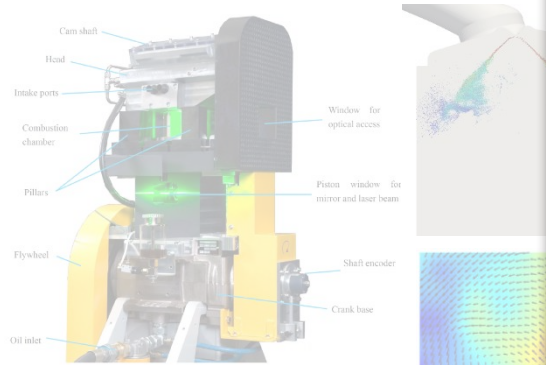
in-cylinder flow



- dynamics of large-scale tumble vortex
- varying field of view
- instantaneous mean field can replace ensemble mean

J. Knöll, E. Mäteling, M. Braun, M. Klaas, & W. Schröder. *Analysis of the in-cylinder flow of a DISI engine using high-speed particle image velocimetry*. LX Laser (2022)

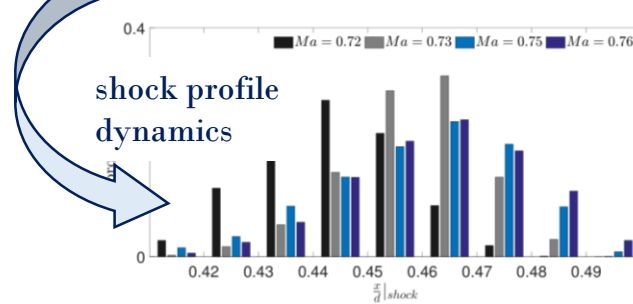
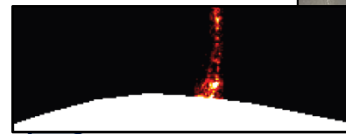
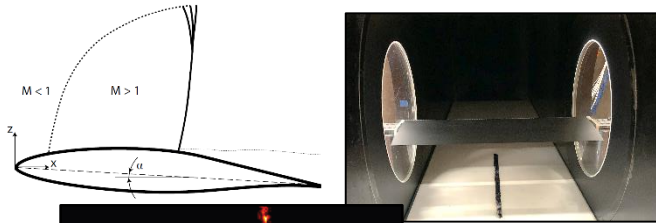
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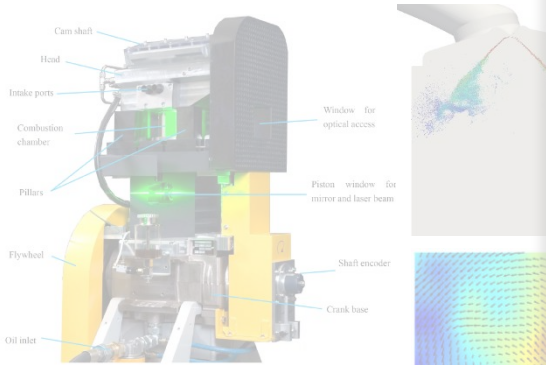
J. Knöll, E. Mäteling, M. Braun, M. Klaas, W. Schröder. *Analysis of the in-cylinder flow in a DISI engine using high-speed particle image velocimetry*. LX Laser (2022)

transonic airfoil buffet



C. Lagemann, E. Mäteling, M. Klaas, & W. Schröder. *Analysis of PIV Images of Transonic Buffet Flow by Recurrent Deep Learning Based Optical Flow Prediction*. LX Laser (2022)

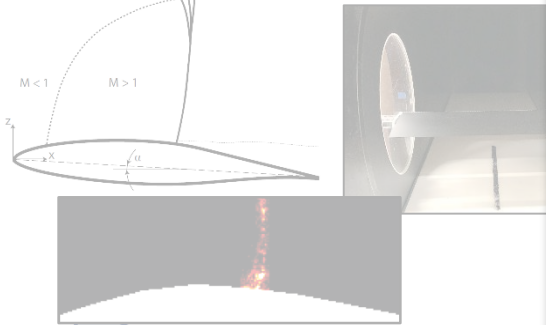
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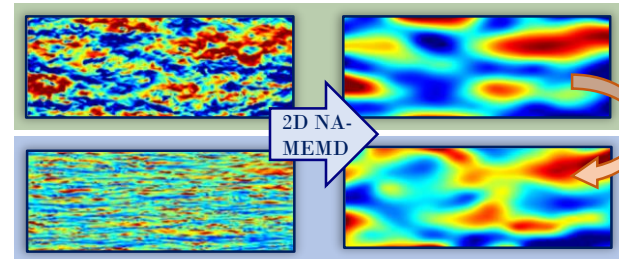
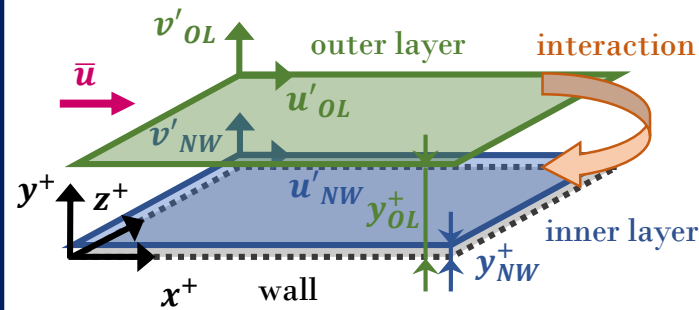
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transonic airfoil buffet



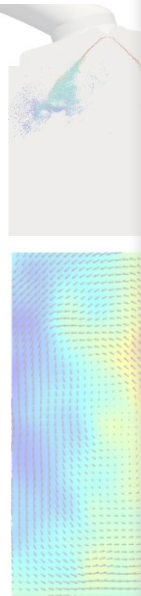
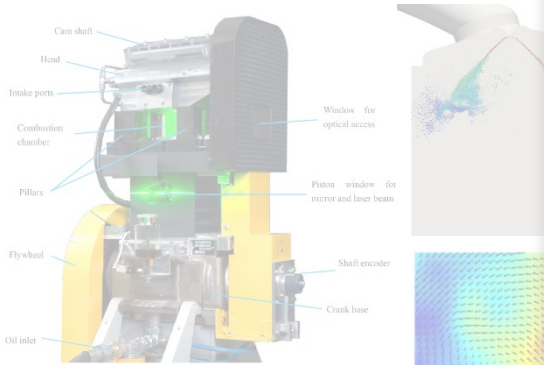
C. Lagemann, E. Mäteling, M. Klaas, W. Schröder. *Analysis of PIV Images of Transonic Buffet Flow by Recurrent Deep Learning for Optical Flow Prediction*. LX Laser (2022)

inner-outer interaction



E. Mäteling & W. Schröder. *Analysis of spatiotemporal inner-outer large scale interactions in turbulent channel flow by multivariate empirical mode decomposition*. Physical Review Fluids 7.3 (2022), p. 034603

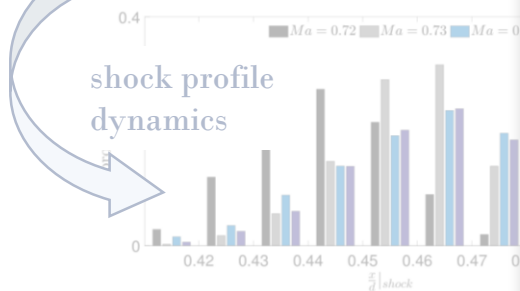
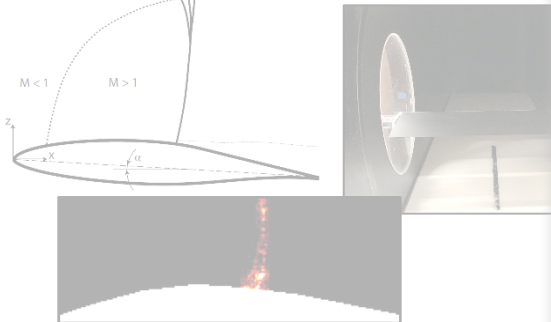
in-cylinder flow



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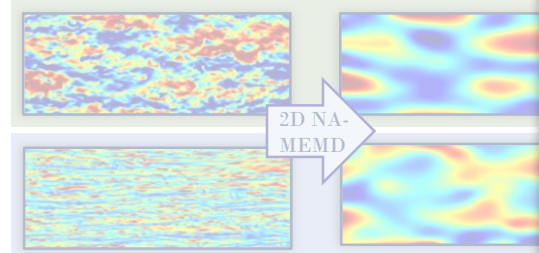
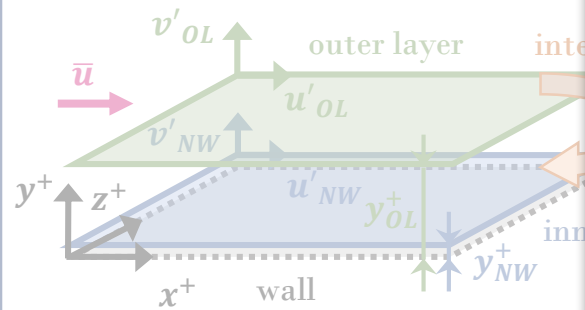
J. Knöll, E. Mäteling, M. Braun, M. Klaas, W. Schröder. *Analysis of the in-cylinder flow in a DISI engine using high-speed particle image velocimetry*. LX Laser (2022)

transonic airfoil buffet



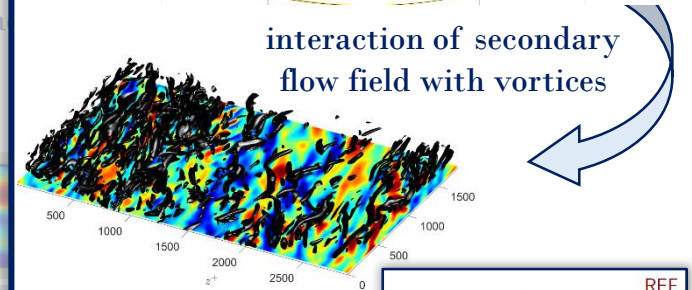
C. Lagemann, E. Mäteling, M. Klaas, W. Schröder. *Analysis of PIV Images of Transonic Buffet Flow by Recurrent Deep Learning for Optical Flow Prediction*. LX Laser (2022)

inner-outer interaction

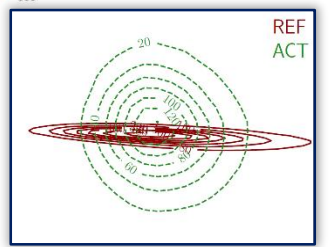


E. Mäteling & W. Schröder. *Analysis of spatiotemporal inner-outer large scale interactions in turbulent flow by multivariate empirical mode decomposition*. Physical Review Fluids 7.3 (2022), p. 033101

drag reduction mechanism



modified inner-outer interaction due to near-wall ejections



E. Mäteling, M. Albers & W. Schröder. *How spanwise travelling transversal surface waves change the near-wall flow*. Journal of Fluid Mechanics 957 (2023), A30



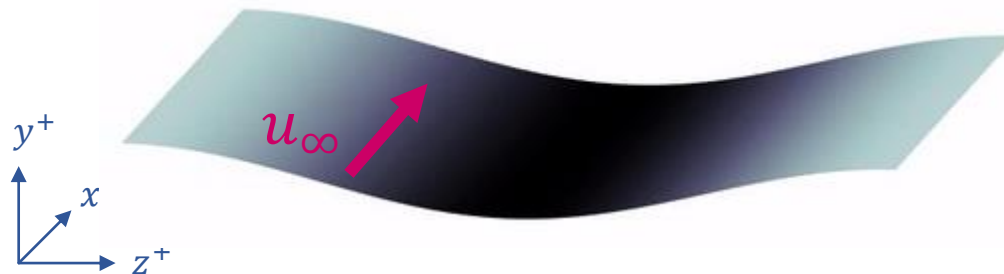
**PHYSICAL MECHANISM BEHIND
ACTIVE DRAG REDUCTION**

How do active drag reduction methods **manipulate the internal flow structure** to achieve drag reduction?

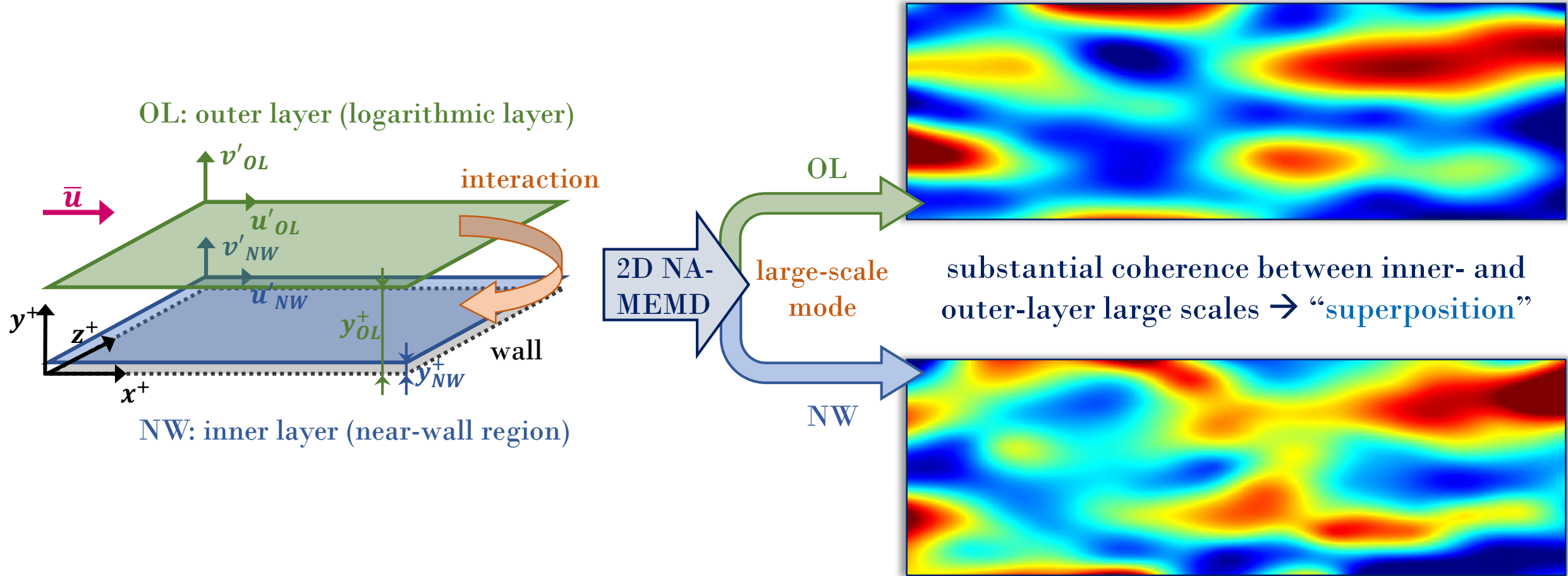
→ tailor methods to any flow configuration with highest efficiency

turbulent boundary layer flow subjected to **spanwise traveling transversal surface waves**

$$y_{wall}^+ = A^+ \cos\left(\frac{2\pi}{\lambda^+} z^+ + \frac{2\pi}{T^+} t^+\right) \quad \text{with } A^+ = 100, \lambda^+ = 3000, T^+ = 50$$

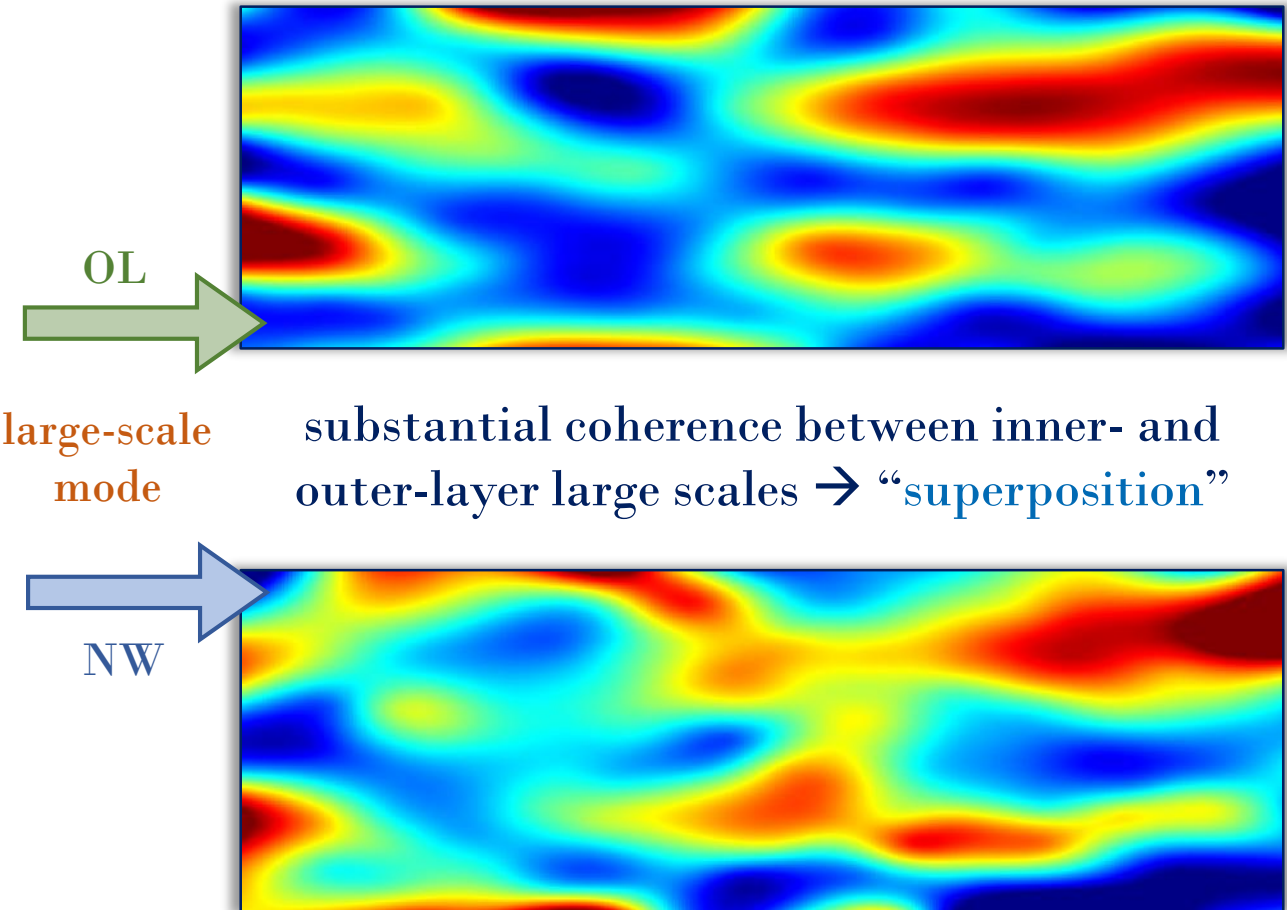
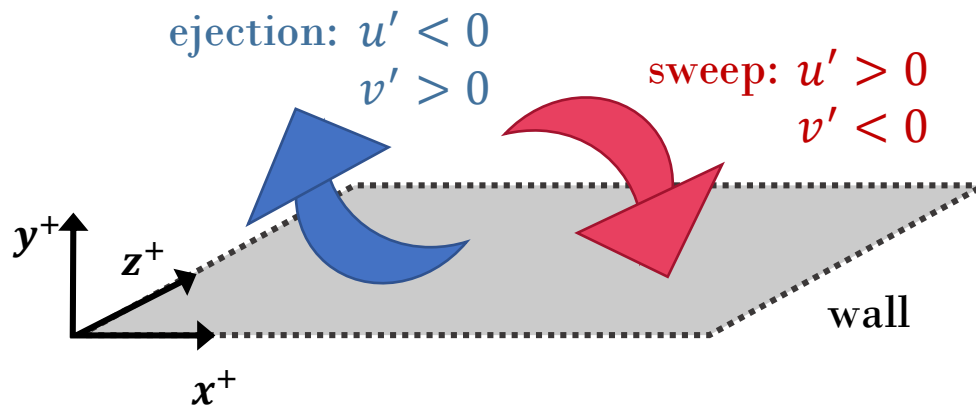


$$\text{Re}_\tau \approx 1500: \quad \Delta c_d \approx 26.5\%$$



interaction phenomena:

- superposition (S)
 - interaction via **sweeps** (sw)
 - interaction via **ejections** (ej)
- amplitude modulation
- frequency modulation

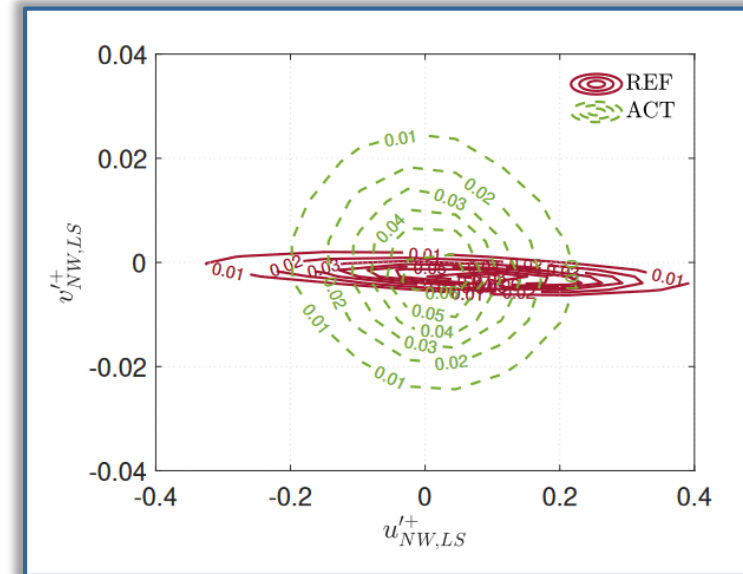


degree of inner-outer coherence
via spatial cross-correlation coefficient

	\bar{R}_s	\bar{R}_{sw}	\bar{R}_{ej}
REF	0.61	0.60	0.36
ACT	0.51	0.51	0.52

- overall, less outer-layer impact on near-wall dynamics (\bar{R}_s)
- reduced top-down communication via sweeps (\bar{R}_{sw})
- increased bottom-up inference related to ejections (\bar{R}_{ej})

joint PDF of large-scale velocity fluctuations close to the wall

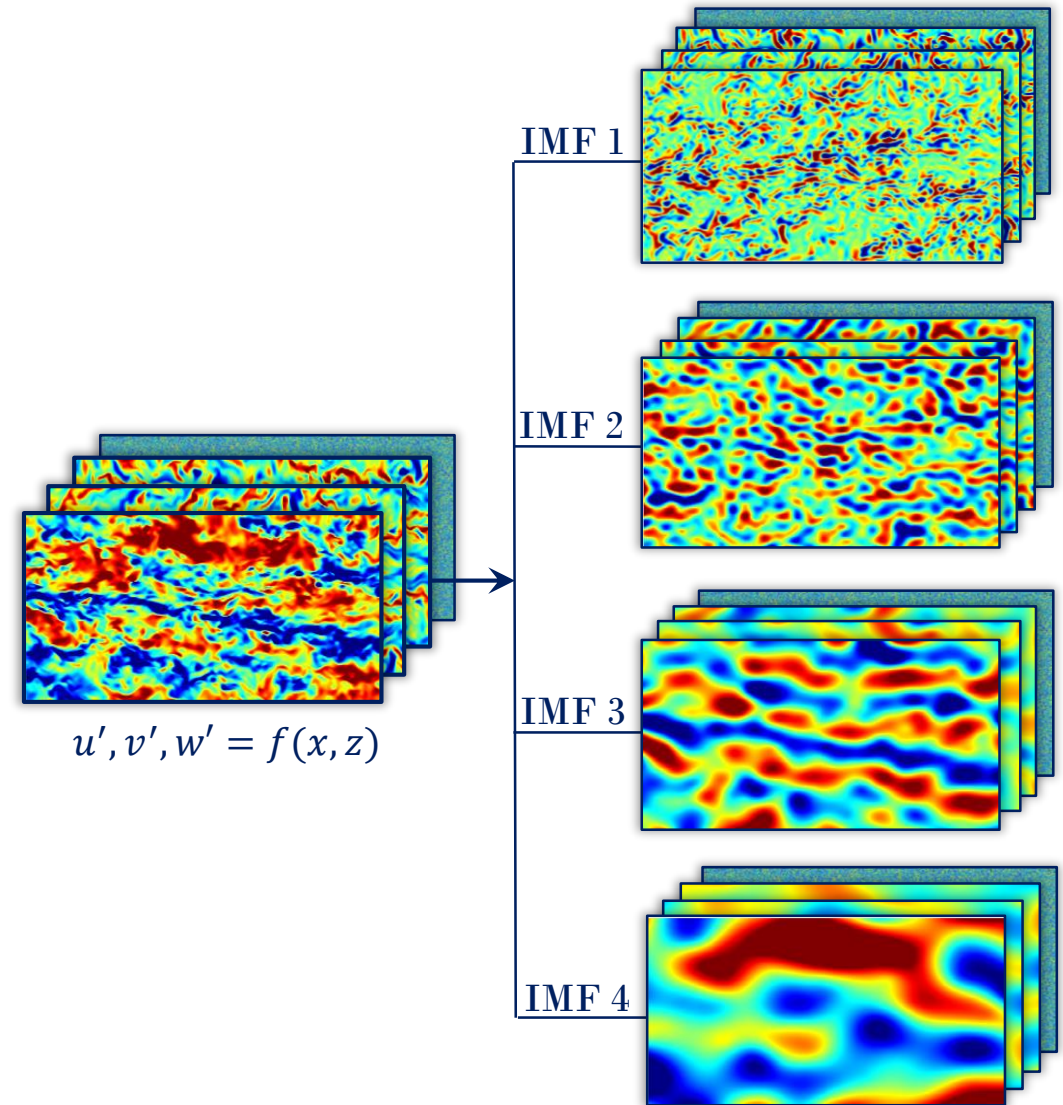


increased number & intensity of near-wall ejections \rightarrow balance outer-layer sweeps
 \rightarrow wall-shear stress reduction



SUMMARY

- introduction of the 2D NA-MEMD for spatio-temporal turbulent flow analyses
 - captures unsteady, non-linear, and multi-scale phenomena
- broad range of applicability
 - numerical & experimental data
 - internal & external flows
 - fundamental turbulence research & real-world/industry-related cases
- example: new insight in physical mechanism behind active drag reduction



**Thank you very much
for your attention!**

Questions?



2D NA-MEMD code

