

**VOLKSWAGEN**

AKTIENGESELLSCHAFT

GROUP INNOVATION

# **REDUCED ORDER MODELS FOR INTERACTIVE AERODYNAMIC VEHICLE DESIGN**

MARKUS MROSEK AND CARSTEN OTHMER, 21.09.2021

VOLKSWAGEN AG

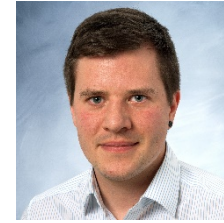


# AGENDA

1. Volkswagen as a Partner in ARIA
2. Interactive Aerodynamic Design: Pre-ARIA Status
3. The Volkswagen ARIA Test Case
4. First ARIA Results of the Hybrid Neural Network/ROM Approach

## PEOPLE

- Markus Mrosek, MSc Mechan. Eng., PhD Candidate at VW since 09/2017



- Carsten Othmer, PhD in Physics, Senior Researcher and Project Lead, at Volkswagen since 2002



- Volkswagen AG, Wolfsburg, Germany
- Volkswagen Group Innovation



## MOTIVATION

**Problem:** Iterative loop between **aesthetic design** and **aerodynamic computation**: 12h up to several days

**Solution:** **Real-time** aerodynamic prediction for **interactive aerodynamic design**

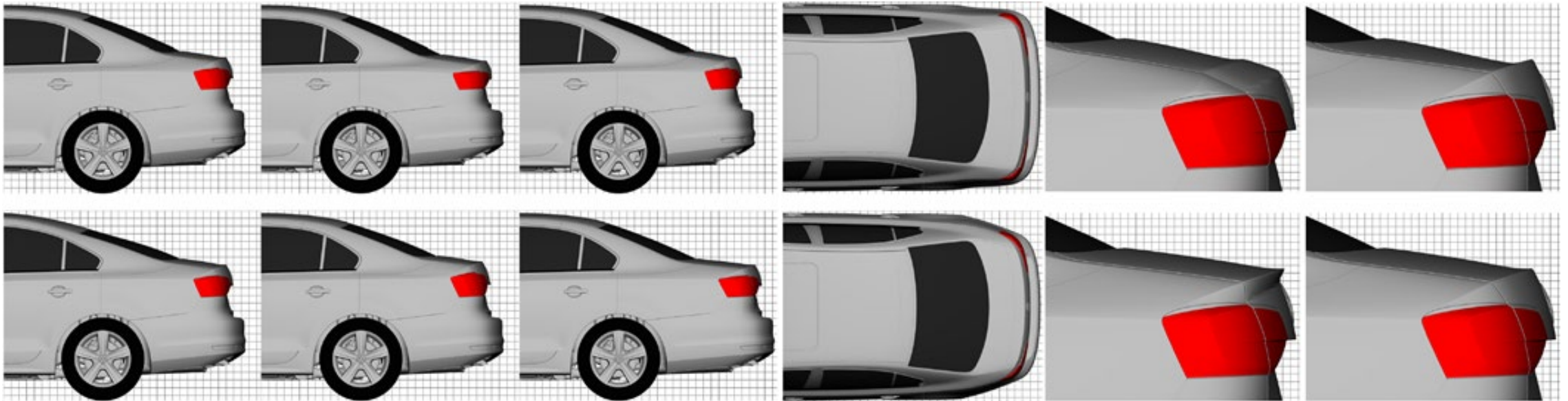
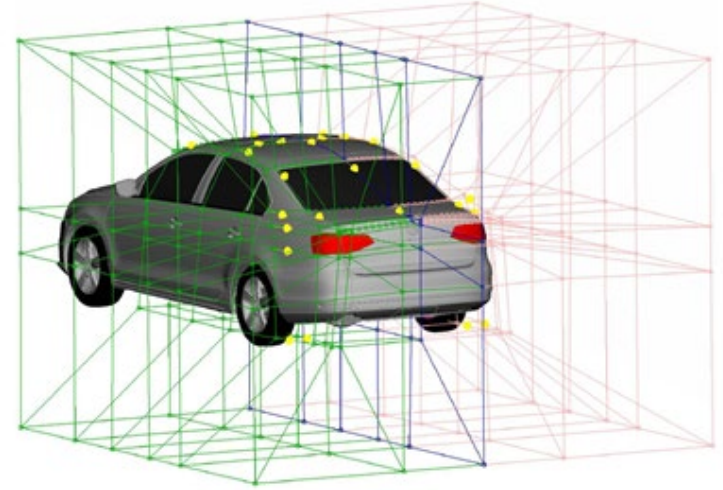


**Required:** Accurate Reduced Order Model (ROM)

# CHARACTERISTICS OF THE REAL-LIFE APPLICATION

## EXAMPLE: 6-DIMENSIONAL JETTA

- ANSA® FFD boxes, 6 parameters in the rear
- $c_d$  and flow fields from OpenFOAM time-averaged DDES
- 103M cells, 4 sec, averaged over the last 2 (= 23,000 core-hours)
- 100 Latin Hypercube samples



a) Rear roof lowering:  
0mm (top), +50mm (bottom)

b) Trunk height:  
-30mm (top), +30mm (bottom)

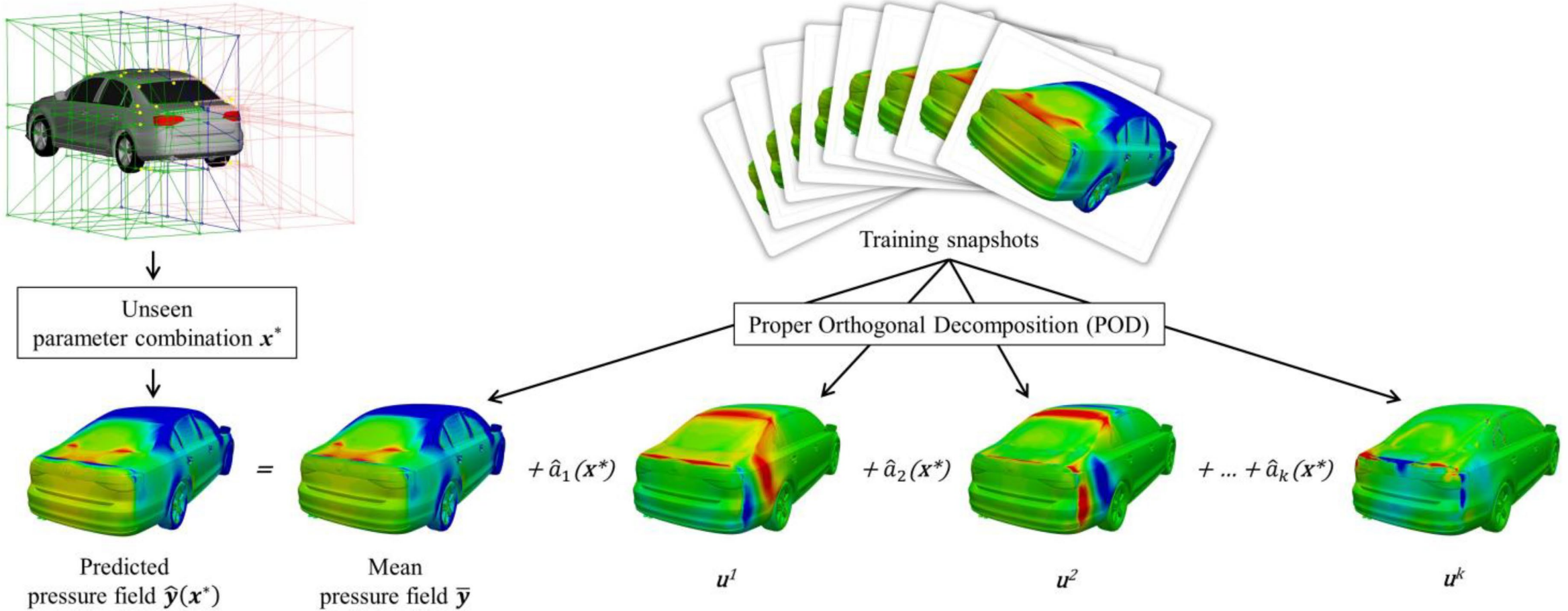
c) Trunk length:  
-50mm (top), +100mm (bottom)

d) Rear lateral tapering:  
-60mm (top), +50mm (bottom)

e) Rear end edge position:  
-70mm (top), +30mm (bottom)

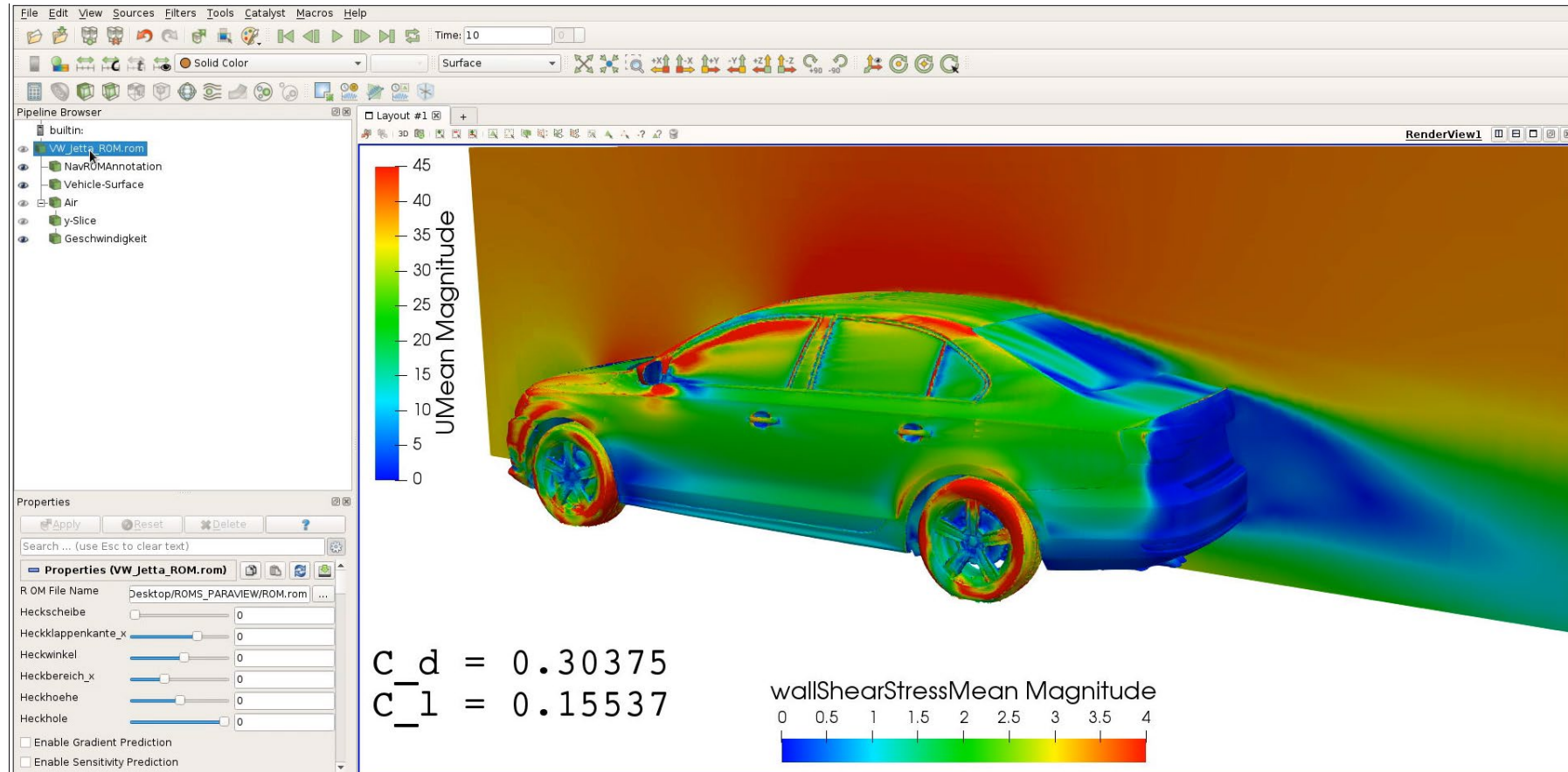
f) Rear end depression:  
-15mm (top), 0mm (bottom)

# CURRENT WORKHORSE: POD + I

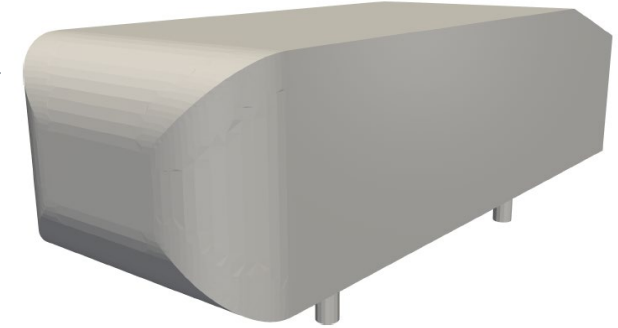


Interpolation of base coefficients with Kriging or Thin-Plate-Splines [SAE-06-12-03-0016]

## HOW IT LOOKS LIKE



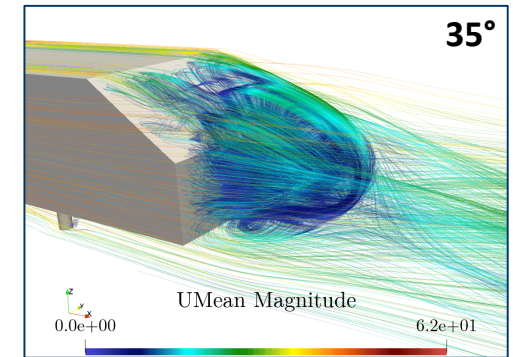
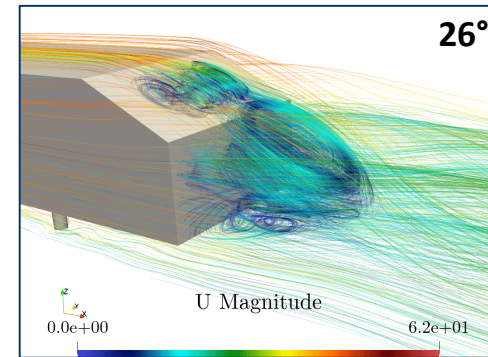
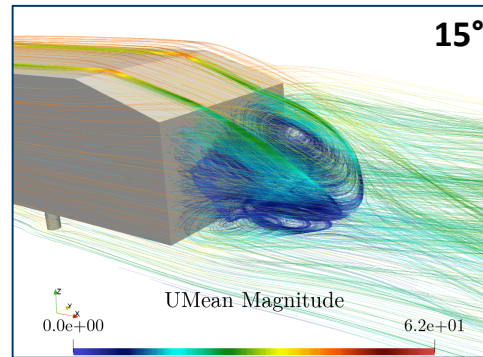
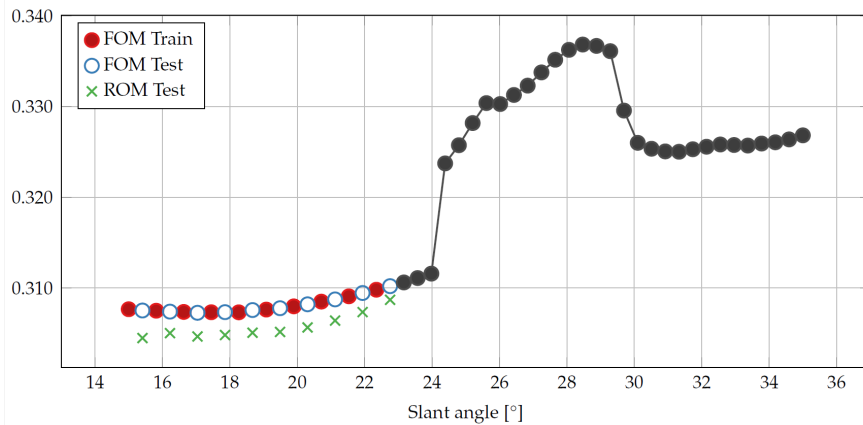
Paraview plugin for interactive visualization (by Navasto, Berlin)



# TEST CASE FOR ARIA: AHMED BODY

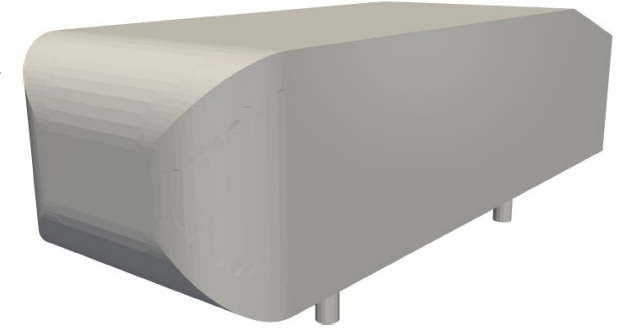
## CASE CHARACTERISTICS

- External aero “classic” [Ahmed et al., 1984], setup created with the SISSA team
- Very coarse on purpose; still physically sensible results



Parameters	Type of parametrisation	CFD-Model	#Cells	#Snapshots	CPU-time / snapshot [core hours]	Storage space / snapshot	Total Storage space
Slant angle: 15-35°	RBF for surface and volume mesh	DES (pimpleFoam)	0.2M	50	16	80MB (1 timestep)	4GB (1 timestep)
		RANS (simpleFoam)	0.2M	50	0.5	80MB	4GB





## TEST CASE FOR ARIA: AHMED BODY

### PREDICTION QUANTITIES AND ERROR METRICS

#### Quantities to be predicted:

- Fields:
  - Wall shear stress and pressure on Ahmed body surface
  - Pressure and vectorial velocity in the volume around the Ahmed body (whole domain)
- Coefficients:
  - Drag coefficient, lift coefficient

#### Error metrics:

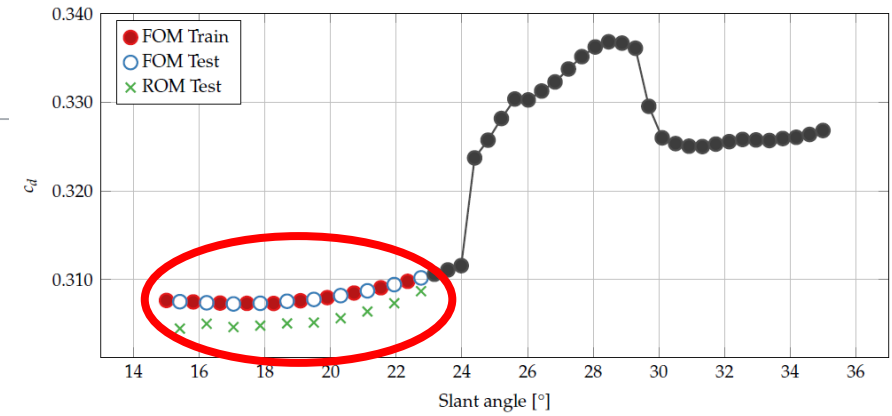
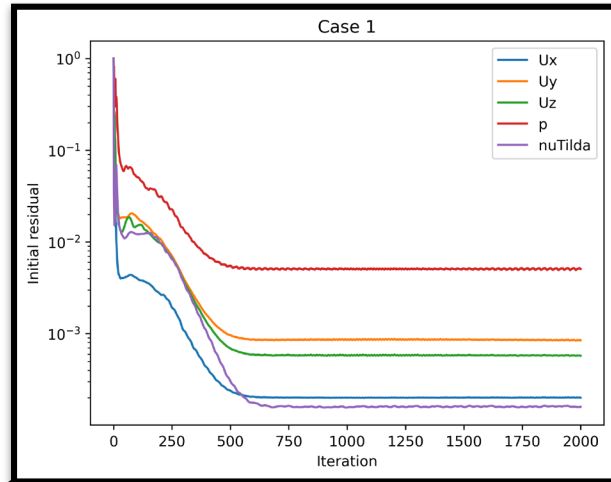
- Fields:  $\delta_y = \frac{\|y - \hat{y}\|_2}{\|y\|_2}$  ( $y$ : CFD field vector,  $\hat{y}$ : ROM field vector)
- Coefficients:  $\delta_c = \frac{|c - \hat{c}|}{|c|}$  ( $c$ : CFD-predicted coefficient,  $\hat{c}$ : ROM-predicted coefficient)

# AGENDA

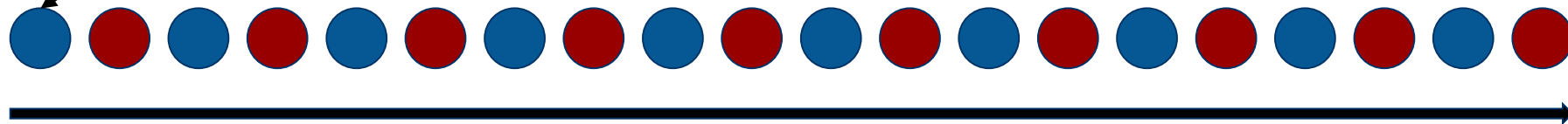
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# VOLKSWAGEN TEST CASE: AHMED BODY

TRAINING DATA - RANS



Train  
Test

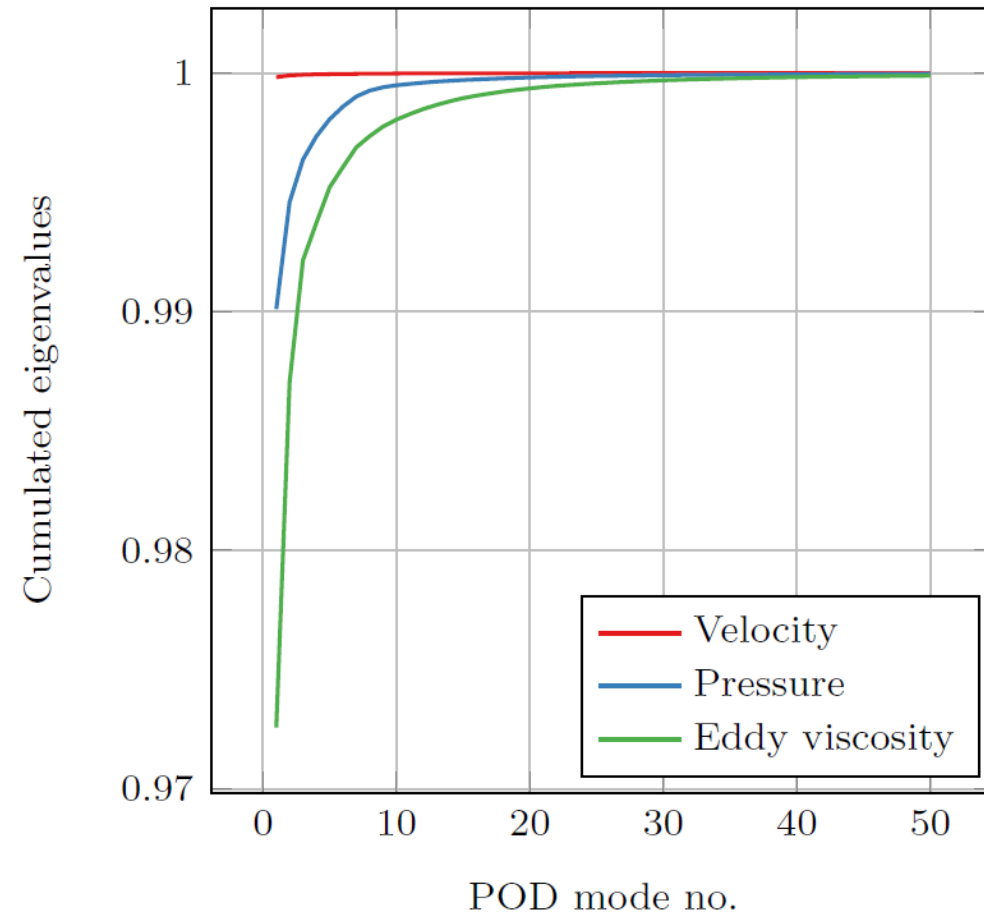


Slant angle: 15 - 22.8 degrees

- 100 snapshots saved per geometry (every 20 iterations out of 2000) → **1000 snapshots**
- Goal is the prediction of converged solution after 2000 iterations

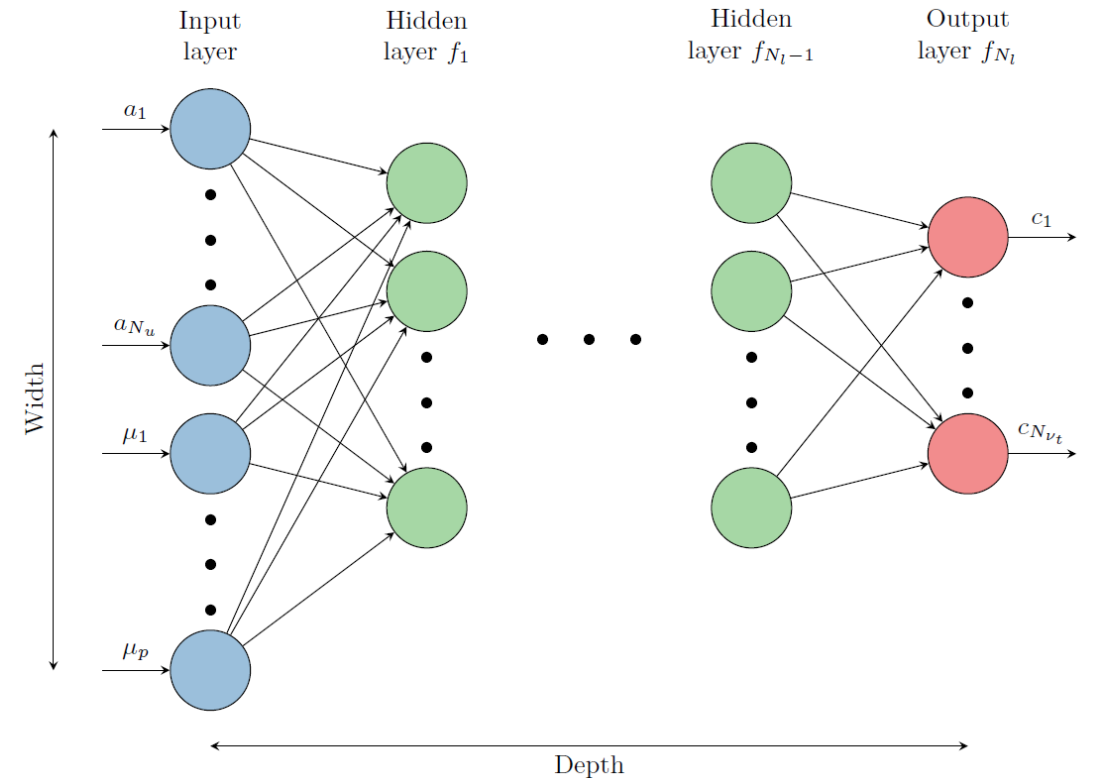
## POD EIGENVALUES

- Separate POD for velocity, pressure and eddy viscosity based on 1000 snapshots
- 30 modes kept for all three fields



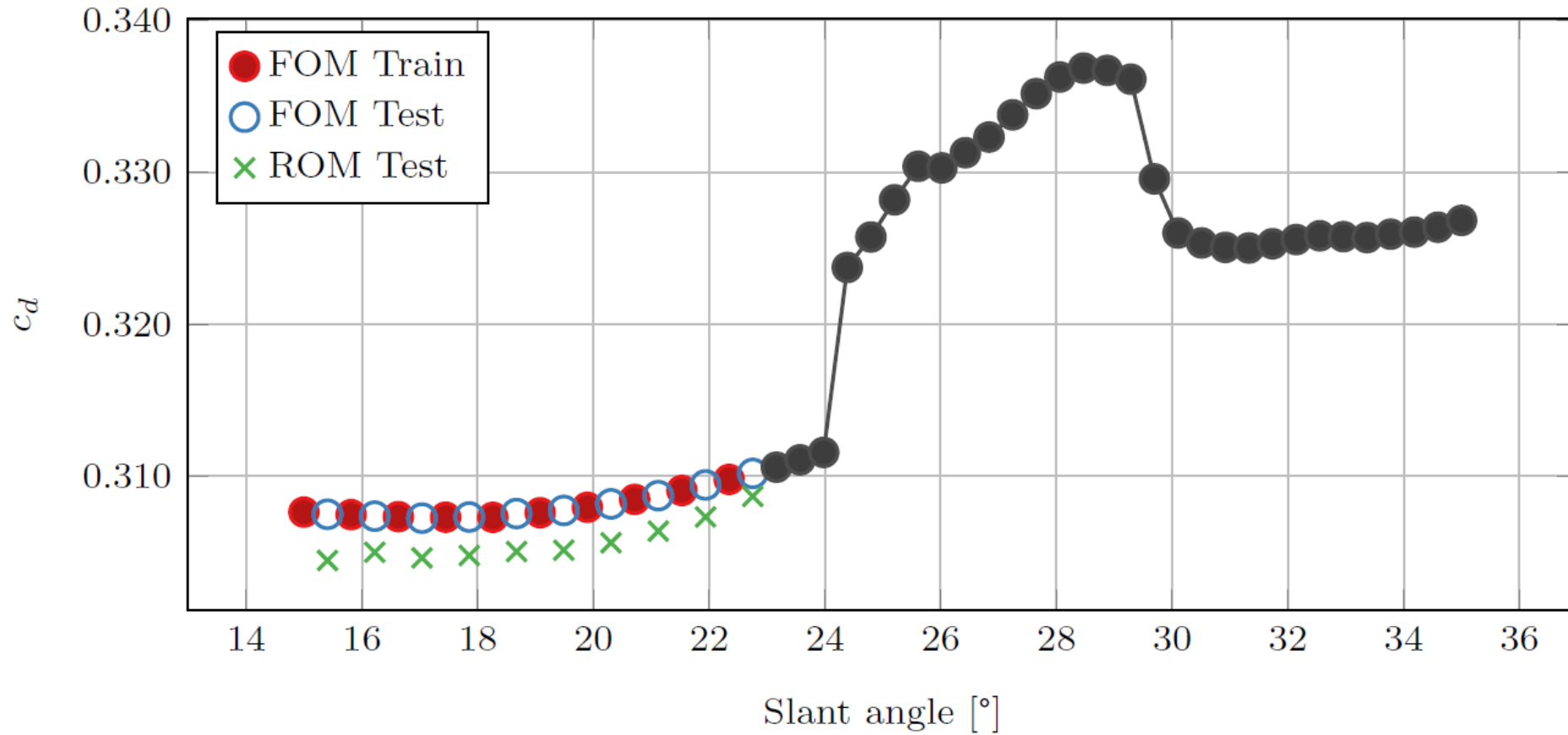
# HYBRID NEURAL NETWORK REDUCED ORDER MODEL

- SIMPLE algorithm for full order model (FOM) and ROM
- Equations for **velocity** and **pressure** are Galerkin projected and then solved
- **Eddy viscosity** modeled by Neural Network
  - Inputs: Slant angle, POD coeffs velocity
  - Outputs: POD coeffs eddy viscosity

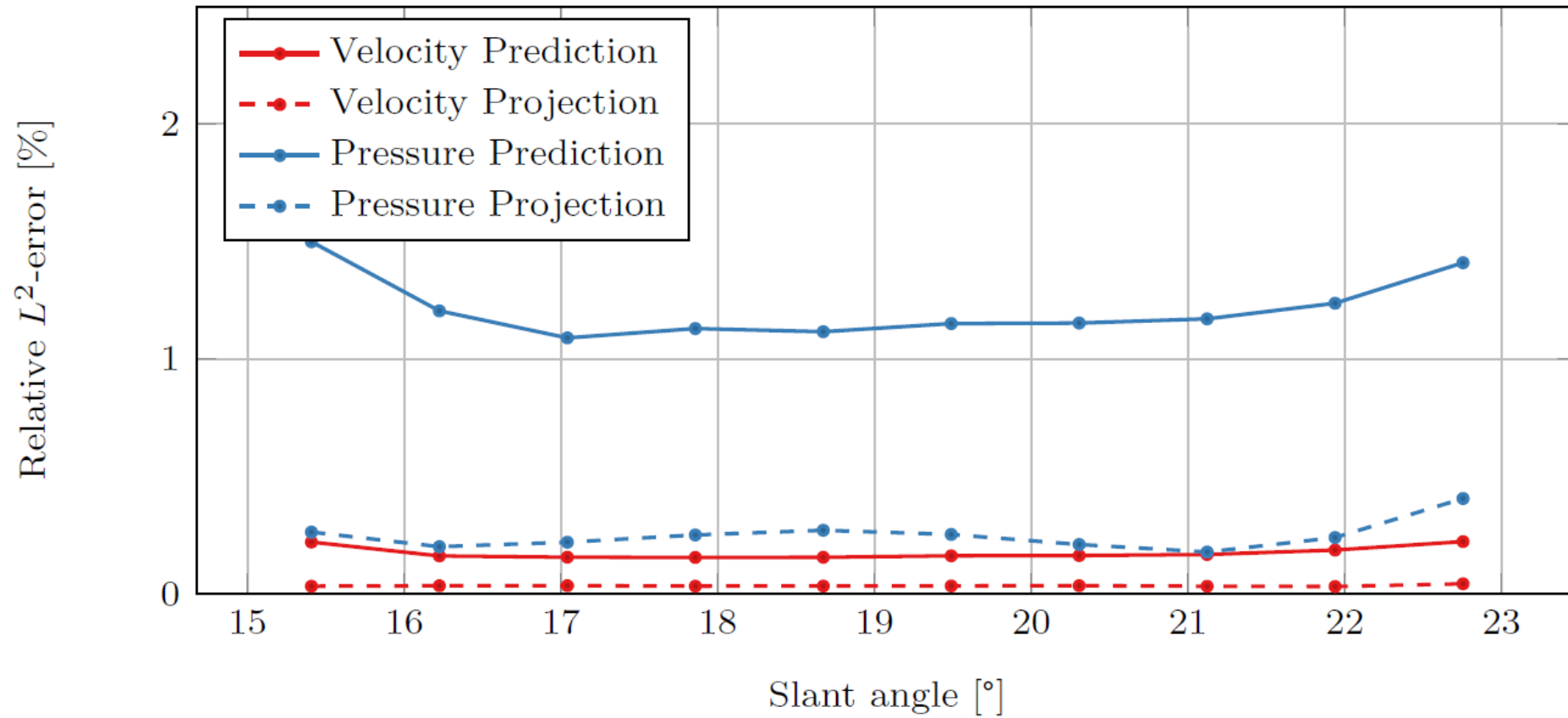


- For details, see Zancanaro et al., Hybrid Neural Network Reduced Order Modelling for Turbulent Flows with Geometric Parameters, Fluids 2021

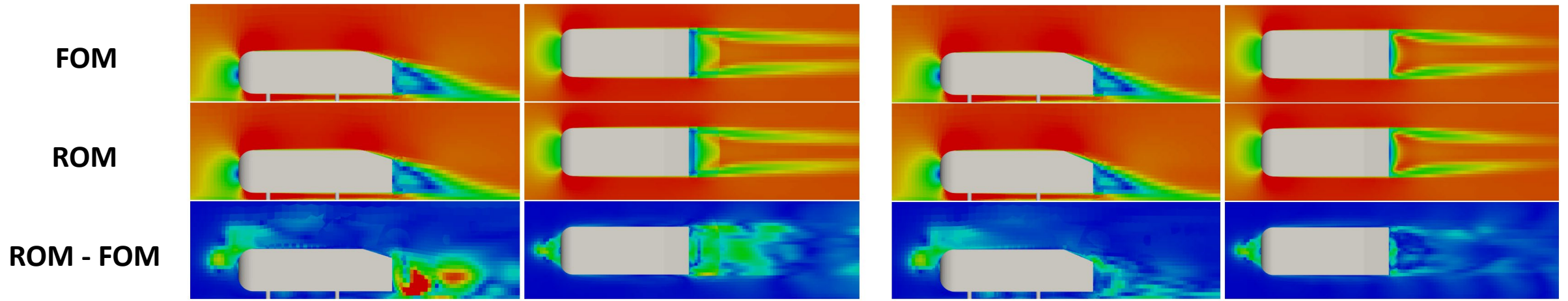
# DRAG COEFFICIENT PREDICTIONS



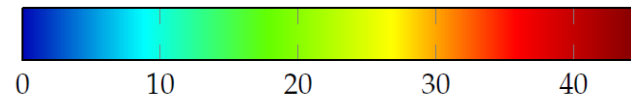
## FIELD PREDICTIONS



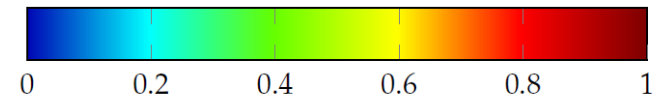
# VELOCITY FIELD PREDICTIONS



Velocity Magnitude [m s<sup>-1</sup>]

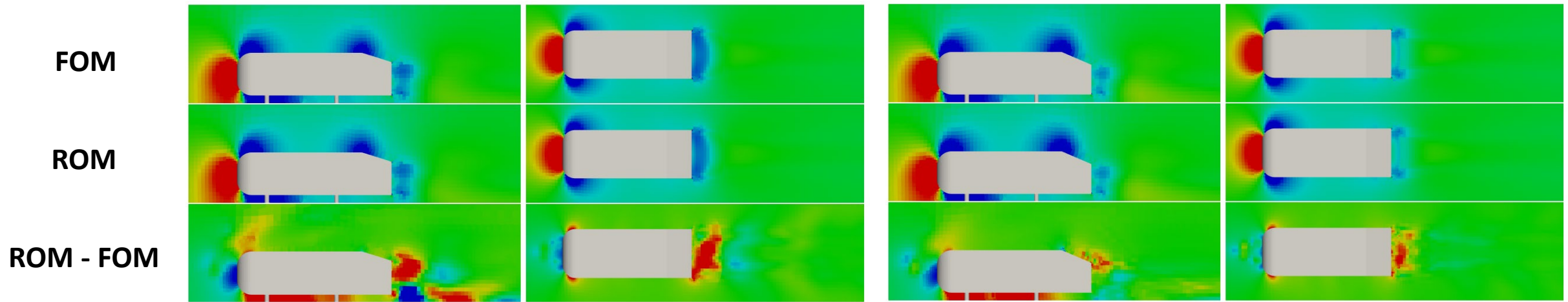


Velocity Difference Magnitude [m s<sup>-1</sup>]





# PRESSURE FIELD PREDICTIONS



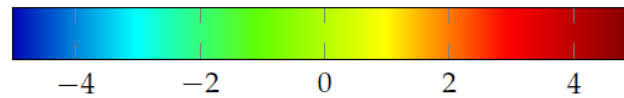
(a) Test sample with slant angle 15.4°.

(b) Test sample with slant angle 22.8°.

Pressure [Pa]



Pressure Difference [Pa]



## SUMMARY AND FUTURE WORK

### Summary:

- Hybrid Neural Network Reduced Order Model successfully applied to Ahmed body test case
- Accurate predictions for velocity and eddy viscosity, while pressure field predictions still show deficiencies

### Future work:

- Improve accuracy of pressure field predictions
- Hyper reduction to decrease prediction times
- Move from RANS to DES

