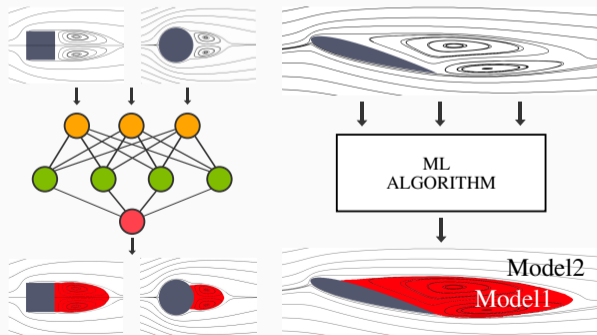


A proposal for automated turbulence modeling

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AIFLUIDs - Chania, May 29 2025



RANS models limitations

Eddy viscosity models are the most used for RANS closure

$$\overline{u'_i u'_j} = -\nu_t \frac{\partial \bar{u}_i}{\partial x_j}$$

Eddy viscosity models:

- resemble constitutive relation of laminar flow
- lack physical ground and generality
- are contaminated by the quest for universality

Prevailing approach

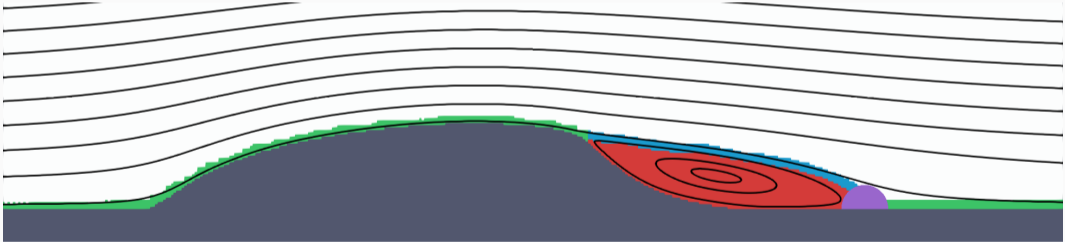
- ML as **alternative** to existing models
- Model PDE are avoided
- Prediction of Reynolds **stresses**
- High fidelity dataset needed
- Driven by data

Our approach

- ML as **complement** to existing models
- Model PDE are used
- Prediction of flow **zones**
- High fidelity dataset not needed
- Driven by physics

Zonal turbulence model

- Modeling each flow zone **independently!**
- Dates back decades ago: e.g., Avva, AIAA 1998



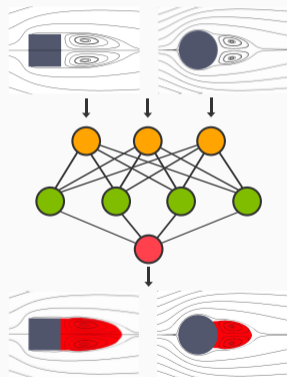
Zonal Automated Turbulence Modeling (ATM)

1. Conceptual design

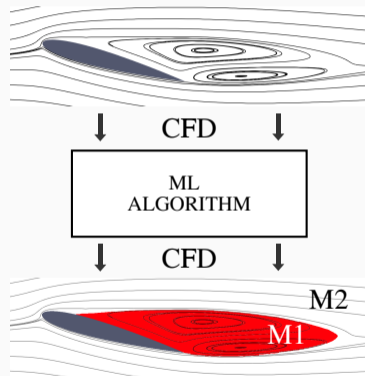
Selection of:

- a) Dictionary of flow zones
- b) RANS model associated to each zone

2. NN for segmentation



3. Zonal ATM



1. Conceptual design: a preliminary implementation

In this work the conceptual design is kept to a **minimum**:

- a) Extremely short dictionary: only two-dimensional flows, three zones
- b) Pure blending of three (existing) models

However, the procedure can be enriched at will.

2. NN for segmentation

ML algorithm for segmentation:

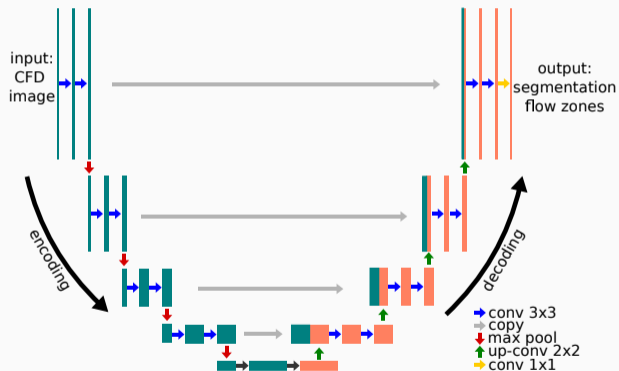
- state-of-the-art **U-net**
- fully convolutional NN
- designed for image segmentation

Input:

- CFD solution as an **image**
- channels with velocity components

Output:

- flow zones segmentation



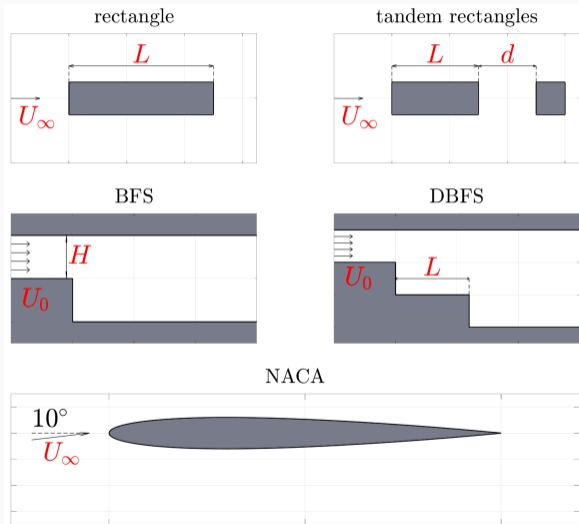
The training dataset

Five classes of 2D steady laminar simulations:

- FreeFem++
- A total of 166 simulations

Dataset generated varying:

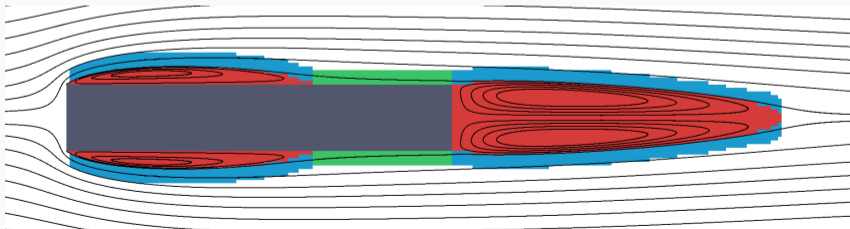
- Reynolds number
- Geometrical features



The labels

Images are labeled **manually**

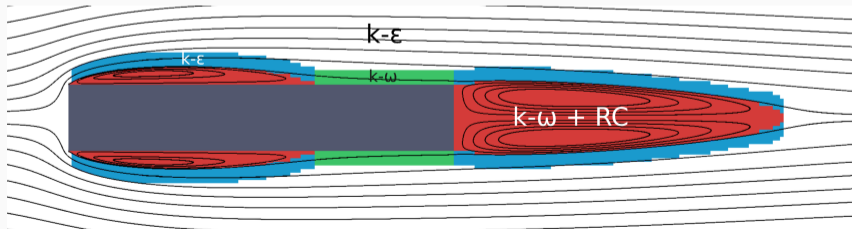
- **Flow recirculation** → stream function
- **Forward attached boundary layer** → velocity and streamlines
- **Shear layer above flow recirculation** → velocity and streamlines
- Solid body (grey)
- Featureless background (white)



3. ATM

Association between zones and RANS models:

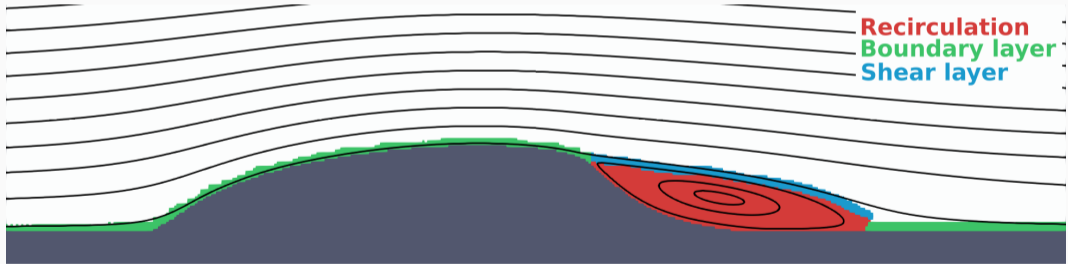
- **Flow recirculation** $\rightarrow k - \omega + \text{Hellsten's Rotation Correction}$
- **Forward attached boundary layer** $\rightarrow k - \omega$
- **Shear layer above flow recirculation** $\rightarrow k - \epsilon$
- Solid body (grey) $\rightarrow -$
- Featureless background (white) $\rightarrow k - \epsilon$



A test case

ATM tested on a **separated** flow: an hump in a channel.

- Reference: experimental data of Greenblatt *et al.* AIAA 2006
- $Re_c = U_0 c / \nu = 936000$, based on hump 'chord' c

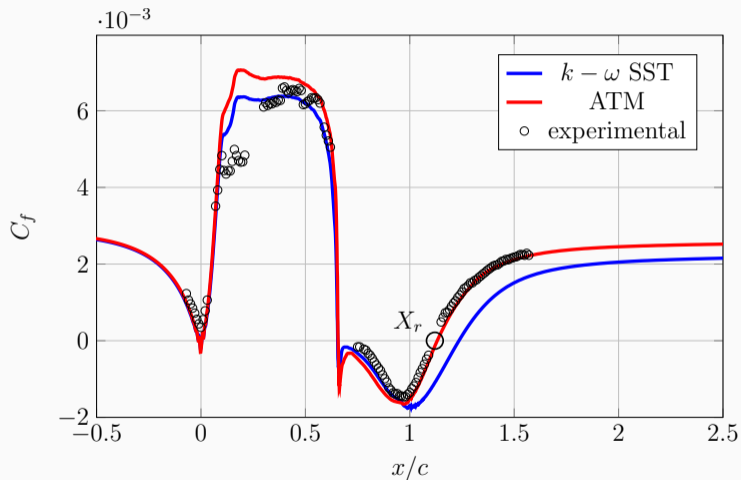


ATM versus established models

model	ΔX_r
$k - \epsilon$	+8.8%
$k - \omega$	+7.3%
$k - \omega$ SST	+11.3%
ATM	+1.9%

ATM versus established models

model	ΔX_r
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ATM	+1.9%



ATM **ticks** all boxes in Spalart (arXiv, 2023):

- ✓ Zonal model without user input
- ✓ **Galilean** invariance
- ✓ **Publishability** of ATM: less than *100Mb* of data
- ✓ Straightforward integration of the NN within a solver

Conclusions

ML **can** be leveraged for a **better use** of existing models:

- Classic eddy viscosity models are **not trashed**
- The segmentation NN is reliable
- Workflow is feasible

However, understanding the **true potential** of the ATM approach requires:

1. Enlarging the dictionary of flow zones
2. Choosing the best model for each zone

Zonal Automated Turbulence Modeling (ATM)

