



## Application of the POD method to optimal design of experiment

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23rd October, 2017

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### Why should we invest in a design of an experiment



- Number of probe locations is always limited.
- For more complex problems intuition is not enough.
- To obtain maximum amount of information from an experiment.
- To reduce measurement error.

1. Perform  $N$  numerical experiments which cover some parameter space (e. g.: different angles of attack, inlet velocities).
2. Create reduced linear model based on numerical data.
3. Use reduced model to construct a statistical criterion.
4. Choose measurement points that will minimize the variance of the model parameters.
5. Perform the experiment and measure the flow parameters in optimal positions.
6. Reconstruct the flow field by using the reduced model and the experiment data.

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## Proper Orthogonal Decomposition I

CFD Simulation  $\rightarrow$  Reduced model  $\rightarrow$  optimization  
 measurement + Reduced model  $\rightarrow$  reconstruction

Requirements:

- Generic.
- Numerically stable.

Reduced model by using the Proper Orthogonal Decomposition:

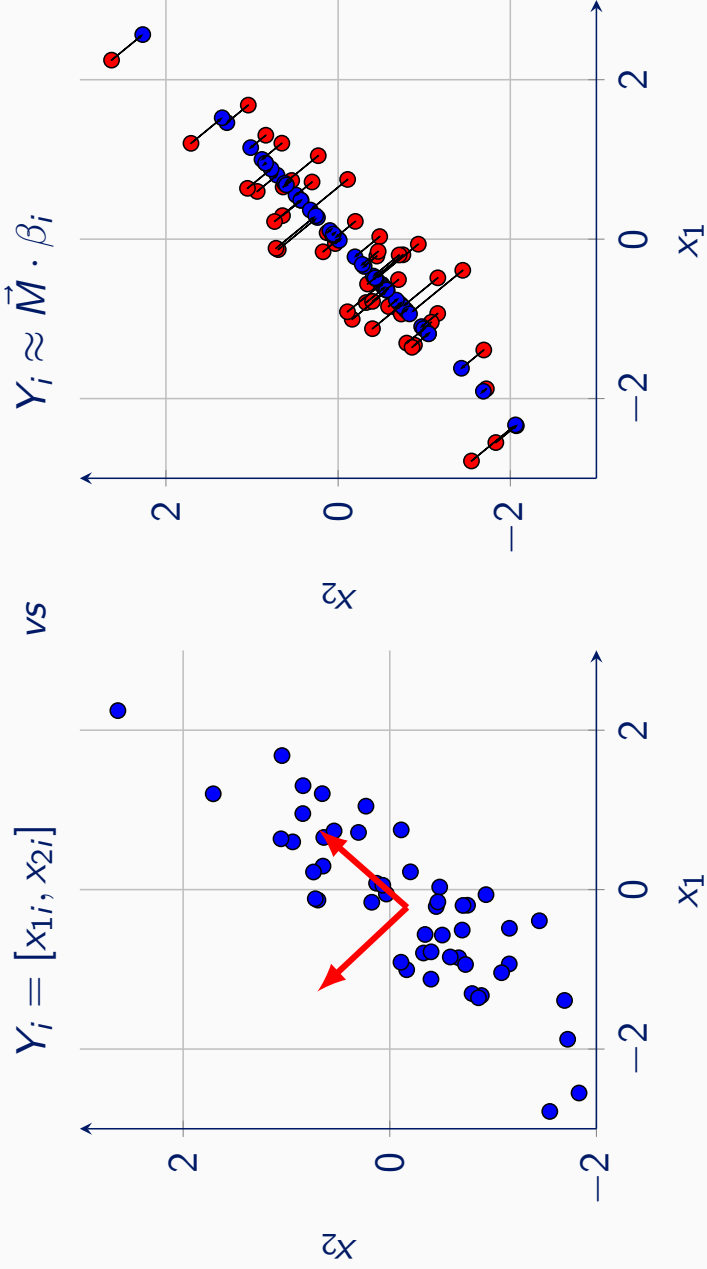
$$\mathbf{V} = [\mathbf{V}_1, \dots, \mathbf{V}_n] \approx \left[ \sum_{i=0}^m \mathbf{M}_i \cdot \beta_{1i}, \dots, \sum_{i=0}^m \mathbf{M}_i \cdot \beta_{ni} \right]$$

$\mathbf{V}_i$  - Data in column vector

$\mathbf{M}_i$  - Basis vectors obtained by the POD

$\beta_i$  - Model coefficients vector

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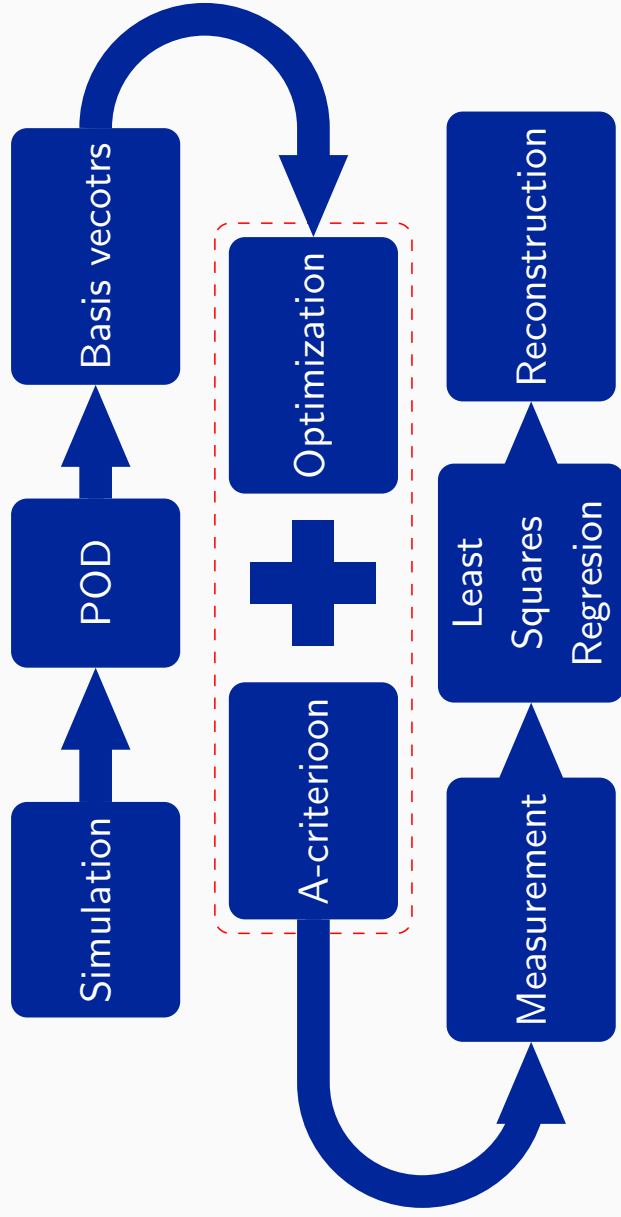
## Statistical criterion

- Linear model describing measured phenomenon:

$$\mathbf{Y} = \mathbf{X} \cdot \beta + \varepsilon$$

- $\mathbf{Y}$  - measurement data at any position that can be predicted by the model,  $\beta$  - model coefficients,  $\mathbf{X}$  - model constants (matrix constructed from POD basis),  $\varepsilon$  - random error.
- Only few equations are required to find  $\beta$ . Selecting an appropriate relations can improve accuracy of the  $\beta$  estimation.
- **A-criterion** - statistical tool allowing to measure variance of estimated parameter  $\beta$  with respect to selected equations (corresponding to measurement positions).

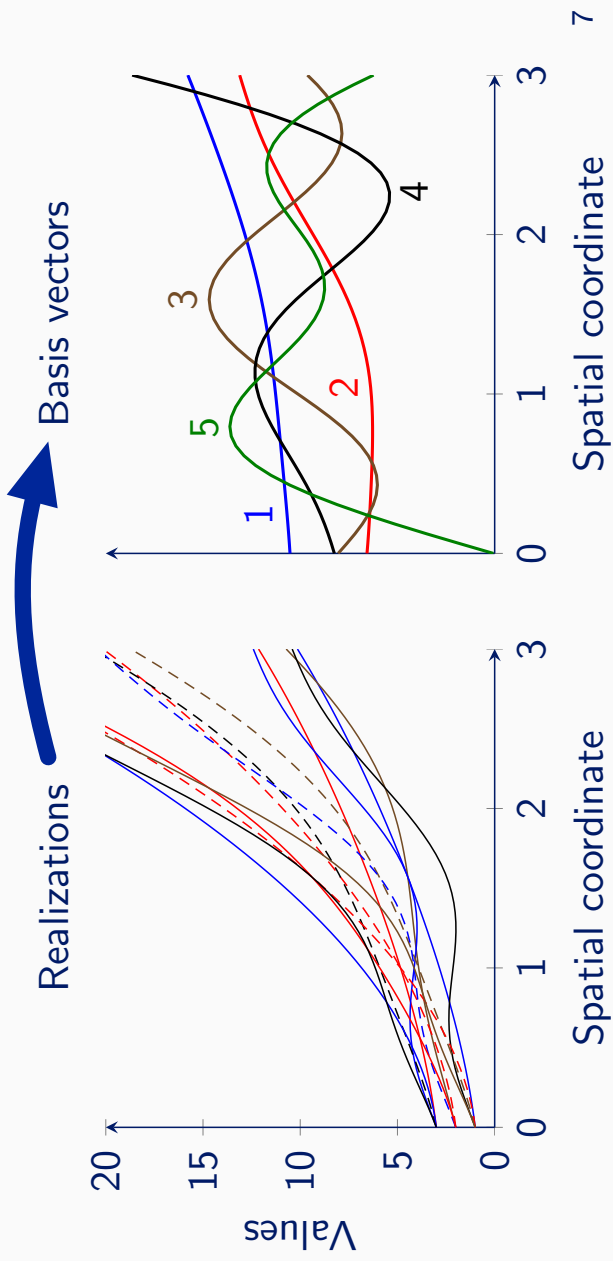
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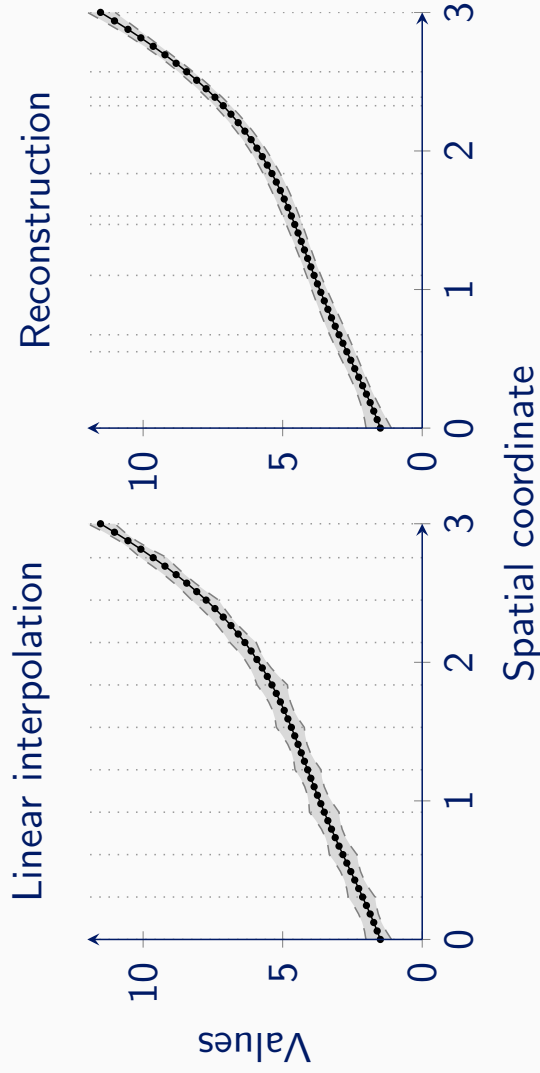
Exemplary case

$$y = a \cdot x^2 + \sin(b \cdot x) + c$$

$$a, b, c = \{1, 2, 3\}$$

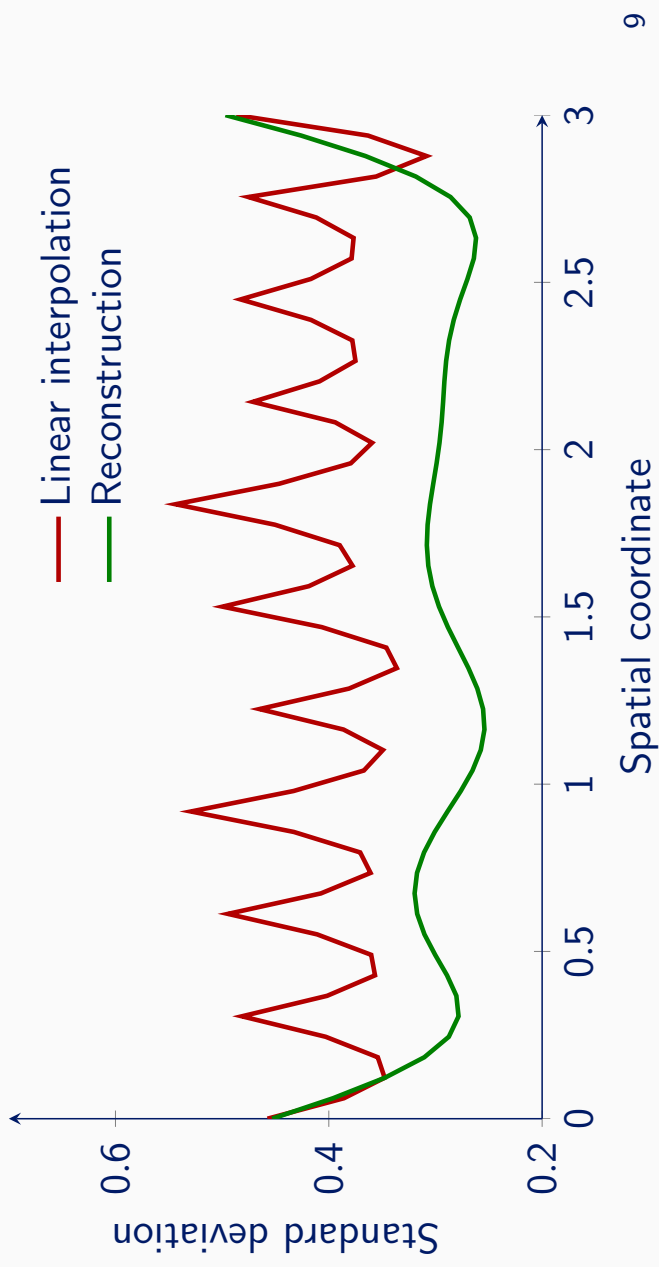


$$a = 1.2 \quad b = 1.8 \quad c = 1.5$$

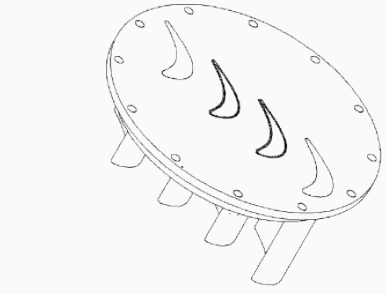


..... Measurement pos.     $\sigma$     —•— Exact

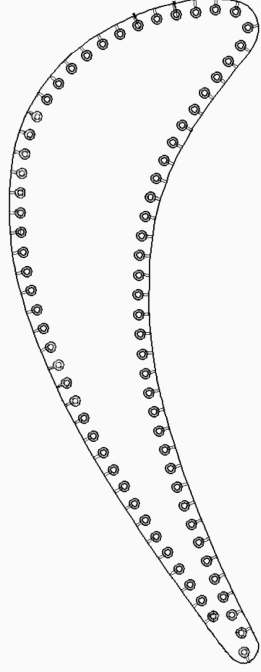
$a = 1.2$        $b = 1.8$        $c = 1.5$



Method validation basing on the experimental data



Experiment - 69 measurement positions on the blade



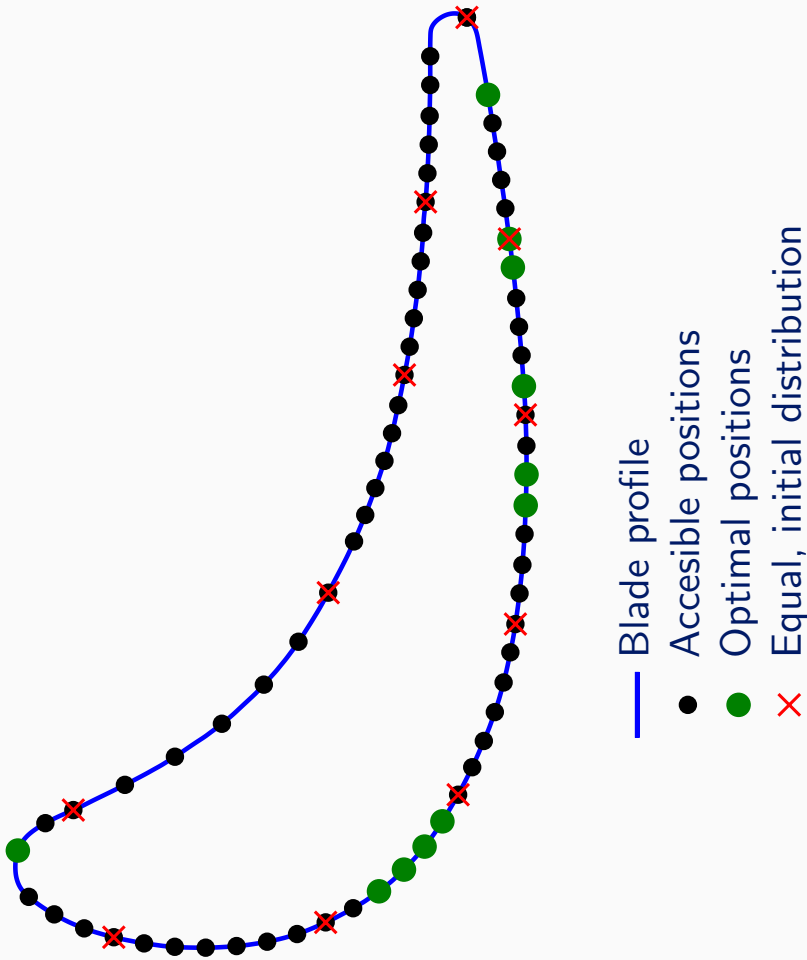
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- Choose 11 optimal out of 69 available positions
- Reduced model constructed (5 basis vectors and parameters) from 200 flow configurations
- Input data for reduced model include variation in angle of attack and inlet Mach number.

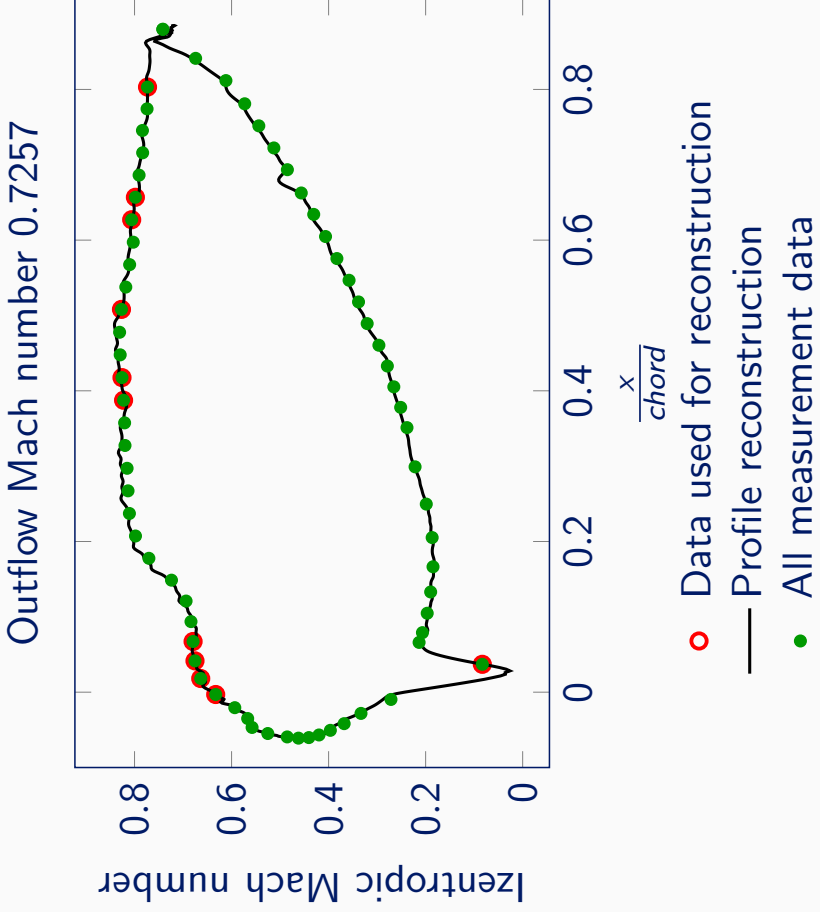
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# Subsonic case

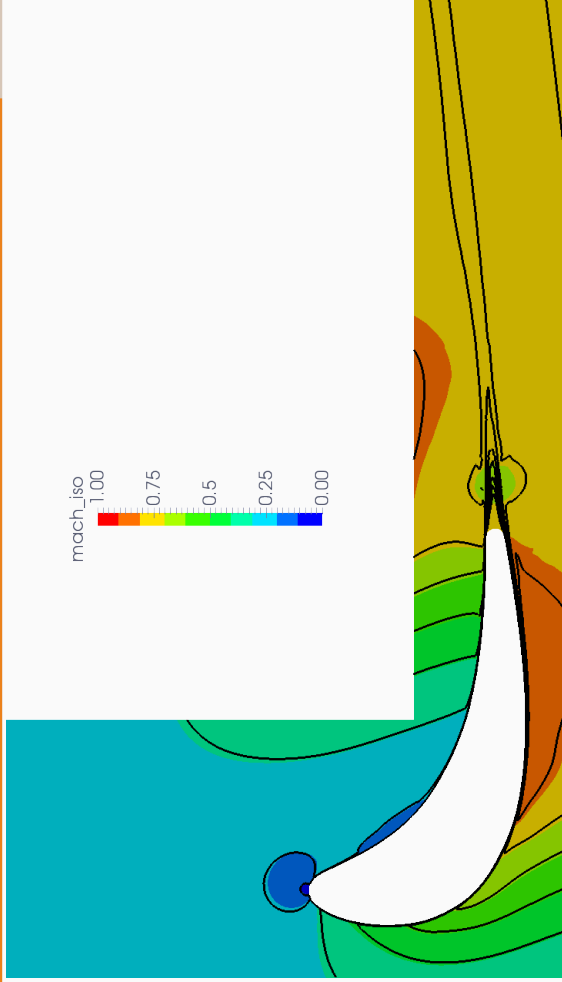
Optimized measurement positions







# Flow field reconstruction

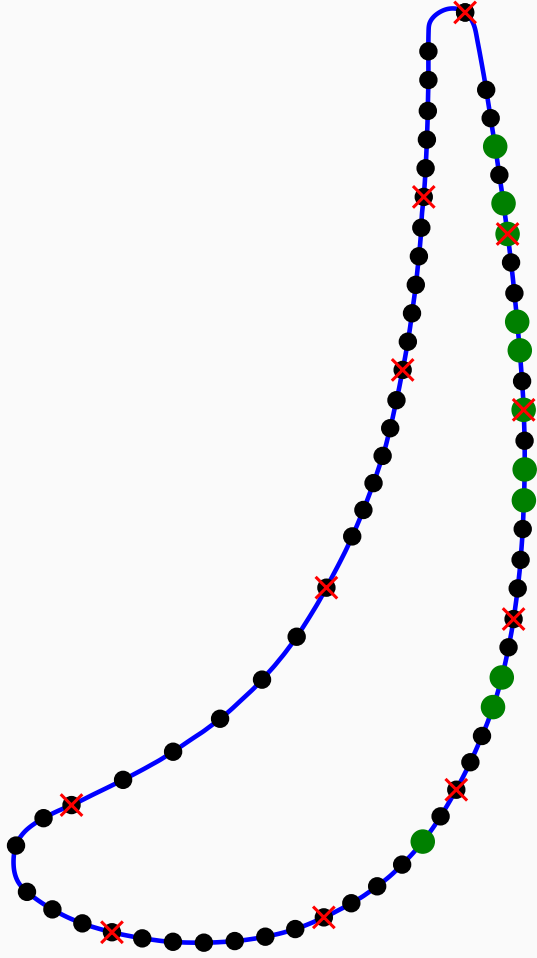


Outflow mach number 0.737

- Filled color - reconstruction
- Black contour lines - closest CFD solution to the experimental conditions

# Sub and supersonic case

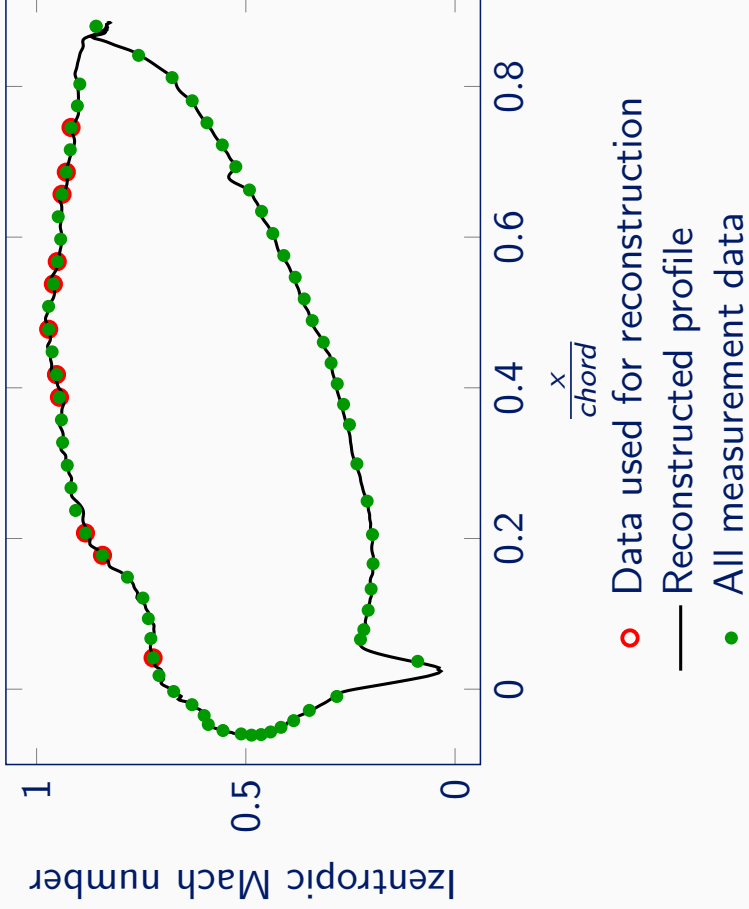
Optimized measurement positions



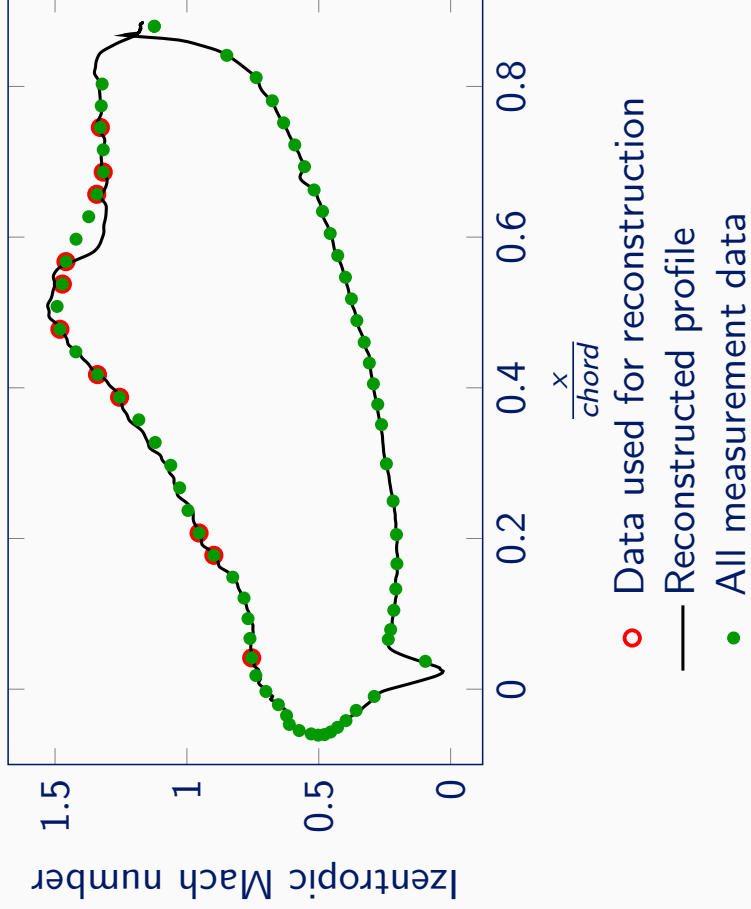
- Blade profile
- Accessible positions
- Optimal positions
- × Equal, initial distribution



Outflow Mach number 0.8398



Outflow Mach number 1.1509





The authors acknowledge support from the research project COOPERNIK financed partly by the Polish National Centre for Research and Development (INNOLOT/I/11/NCBR/2014) and partly by Avio Polska Sp. z o.o.

Thank you for your attention