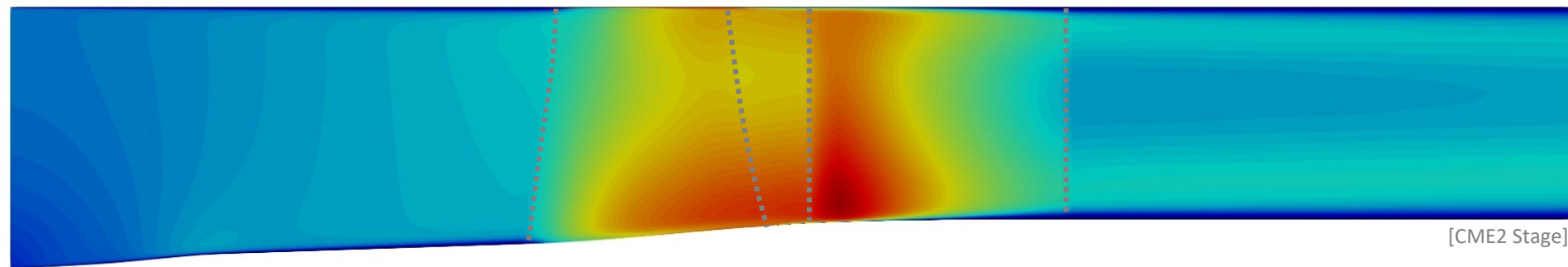


Quantification of geometric variability effects through a viscous through-flow model: sensitivity analysis of the manufacturing tolerance effects on performance of modern axial-flow compressor blades

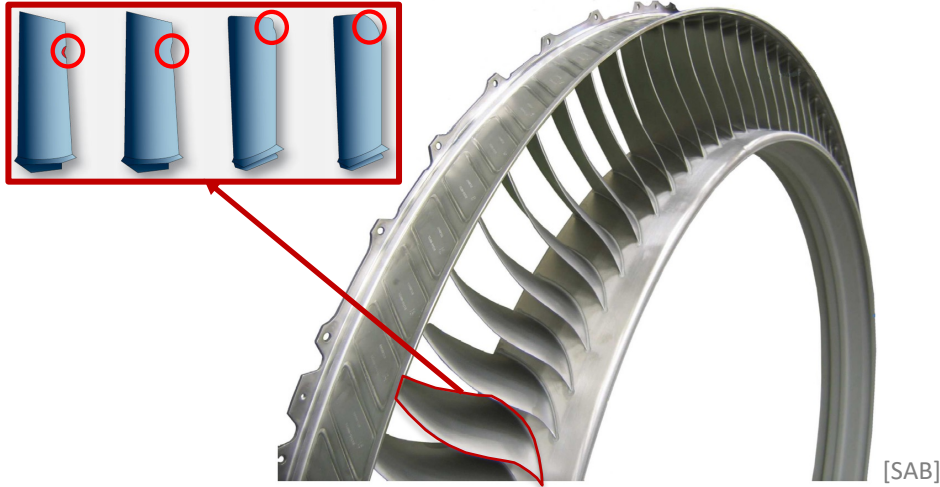


Arnaud Budo⁽¹⁾

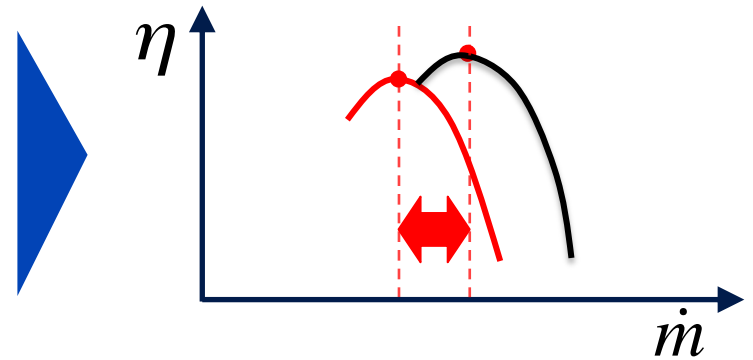
Thibault Le Men⁽²⁾, Koen Hillewaert⁽¹⁾,
Maarten Arnst⁽¹⁾, Vincent E. Terrapon⁽¹⁾

Context

Geometric variability of low-pressure compressor blades



Performance variation



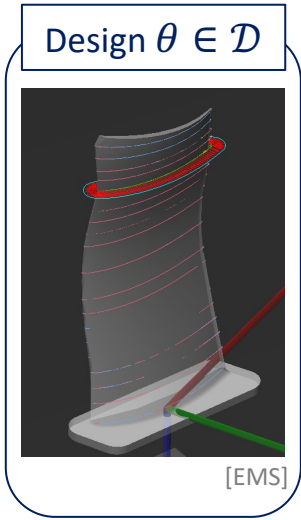
Manufacturing tolerances?

- Need of rigorous/robust definition
- Linked to manufacturing process
- Simplify the treatment of poorly made parts

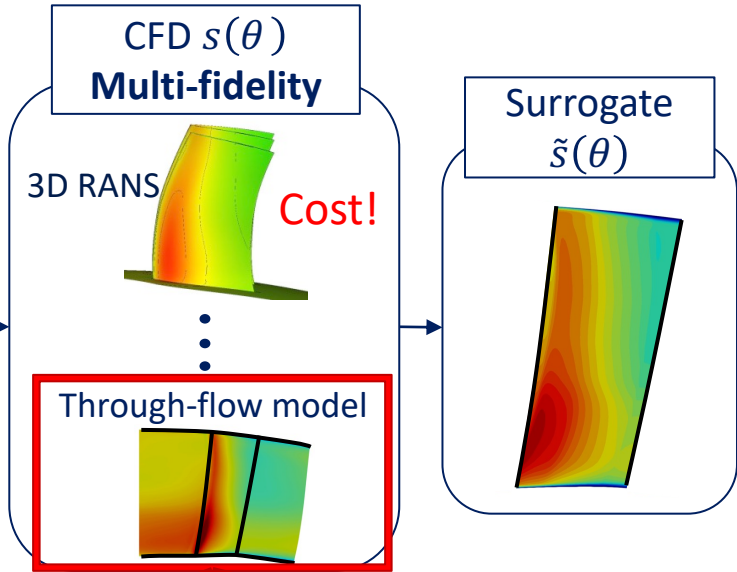
Trade-off
Cost ↔ performance

Methodology & objectives

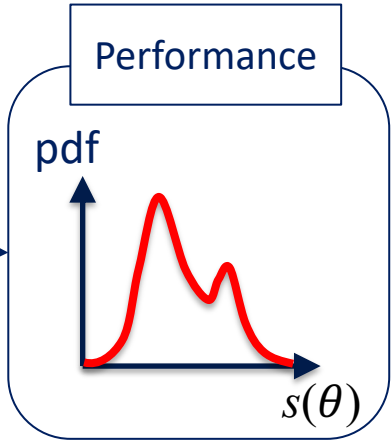
Characterization



Propagation



Qualification



Able to predict performance?

- Through-flow model validation**
- Consistency
 - Choice of model correlations

Able to capture variability effects?

- Geometrical variability**
- Sensitivity analysis
 - (Uncertainty quantification)

Outline

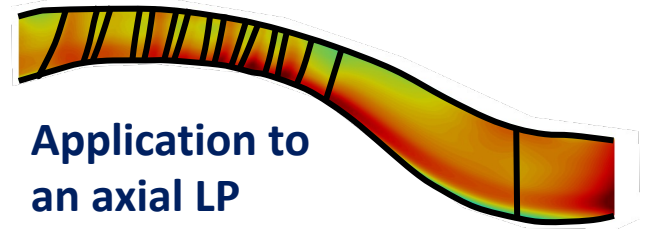
1

**Viscous
through-flow model**

$$\frac{\partial U}{\partial t} + \frac{\partial(F-F_v)}{\partial x} + \frac{\partial(G-G_v)}{\partial r} = S$$

2

Model Assessment



**Application to
an axial LP
compressor**

3

Geometric variability



[SAB]

Outline

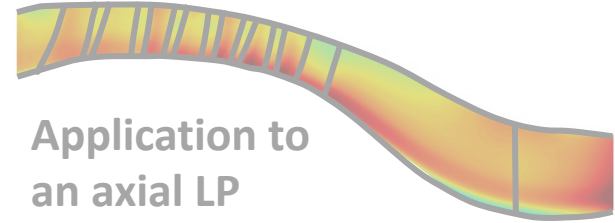
1

**Viscous
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$$\frac{\partial U}{\partial t} + \frac{\partial(F-F_v)}{\partial x} + \frac{\partial(G-G_v)}{\partial r} = S$$

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Model Assessment



Application to
an axial LP
compressor

3

Geometric variability



[SAB]

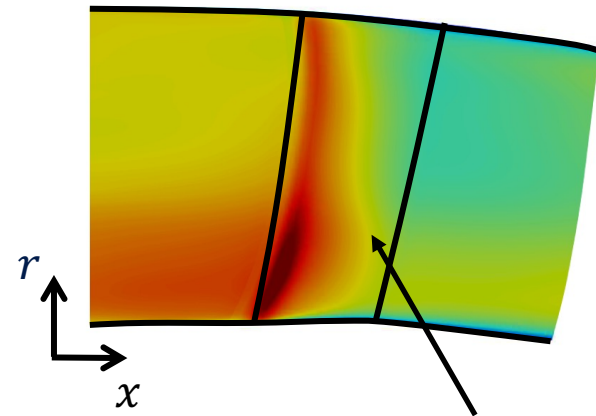
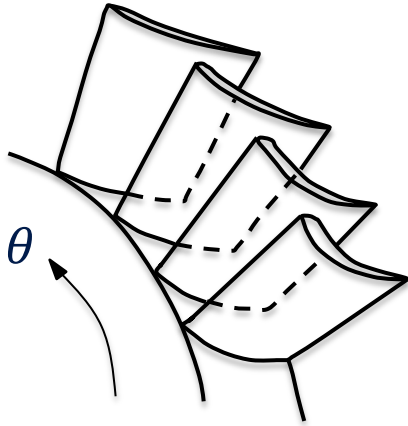
Through-flow model

$$D_t \mathbf{U}(r, \theta, x, t) = \mathbf{G}(\mathbf{U}, r, \theta, x, t)$$

θ -averaging

$$D_t \bar{\mathbf{U}}(r, x) = \bar{\mathbf{G}}(\mathbf{U}, r, x)$$

Unclosed!



Implicit presence of the blades

- Azimuthal/pitchwise averaging approach
- Axisymmetric steady flow (meridional plane)
- Empirical correlations
- Low computational cost $\mathcal{O}(\min)$

Through-flow formulation

$$D_t \bar{U}(r, x) = \bar{G}(U, r, x)$$

Governing equations

- Underlying assumptions (NSE, Euler, SLC, ...)?
- Consistency?

Closure models

- Exhaustive?
- Assumptions?
- Correlations?

Choices influence **level of empiricism** and determine **error sources**

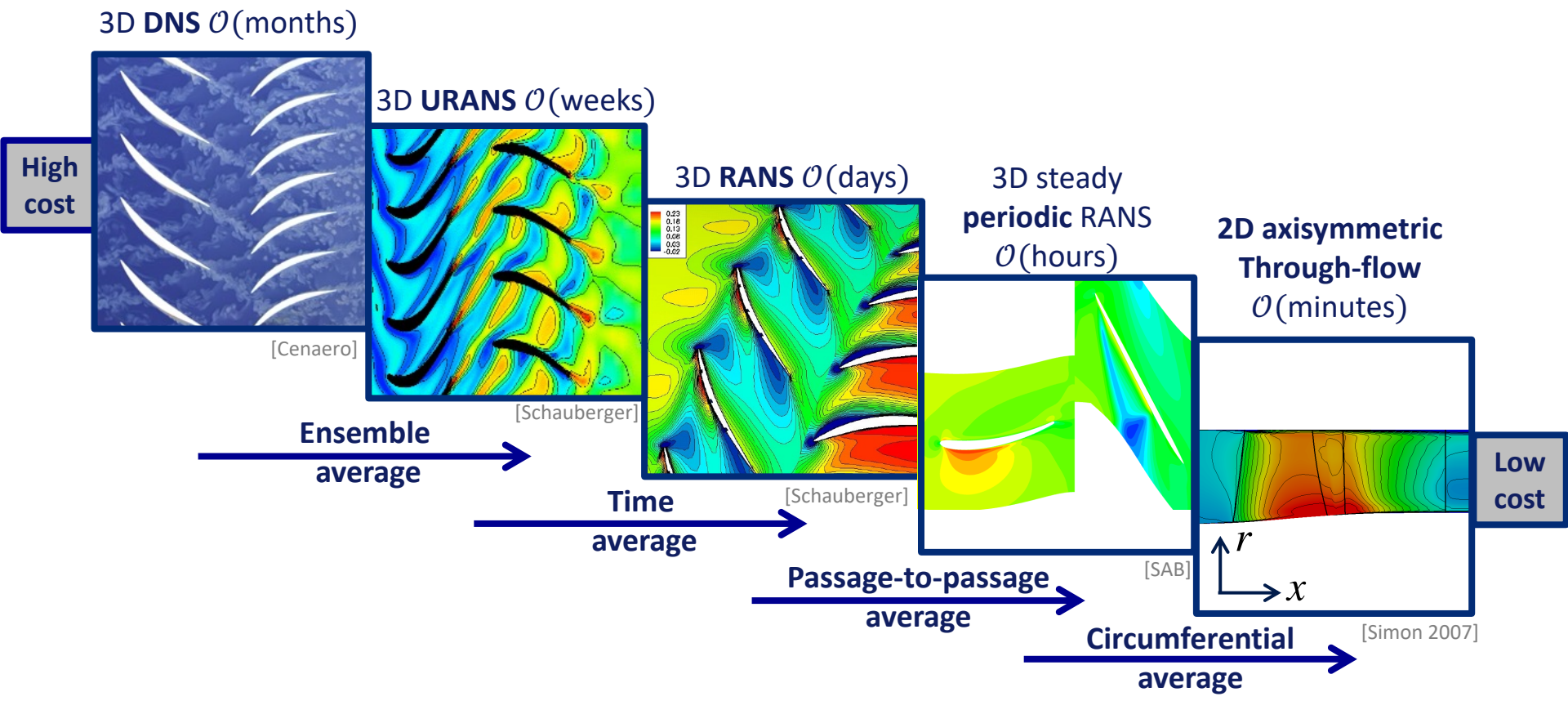
→ **characterize error in a rigorous and exhaustive way**

Adamczyk's cascade

$$D_t \bar{U}(r, x) = \bar{G}(U, r, x)$$

- Exact mathematical formulation of source terms
- NSE-based equations

Unclosed!



Adamczyk's cascade: unclosed terms

$$D_t \bar{U}(r, x) = \bar{G}(U, r, x)$$

Unclosed terms:

$$\overline{\rho V'_i V'_i}$$

Non-linear
equations

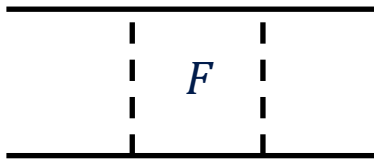
τ_{reys} ➤ Reynolds

τ_{uns} ➤ Unsteady

τ_{ape} ➤ Aperiodic

τ_{circ} ➤ Circumferential

Stresses



B_i ➤ inviscid

B_v ➤ Viscous

Forces

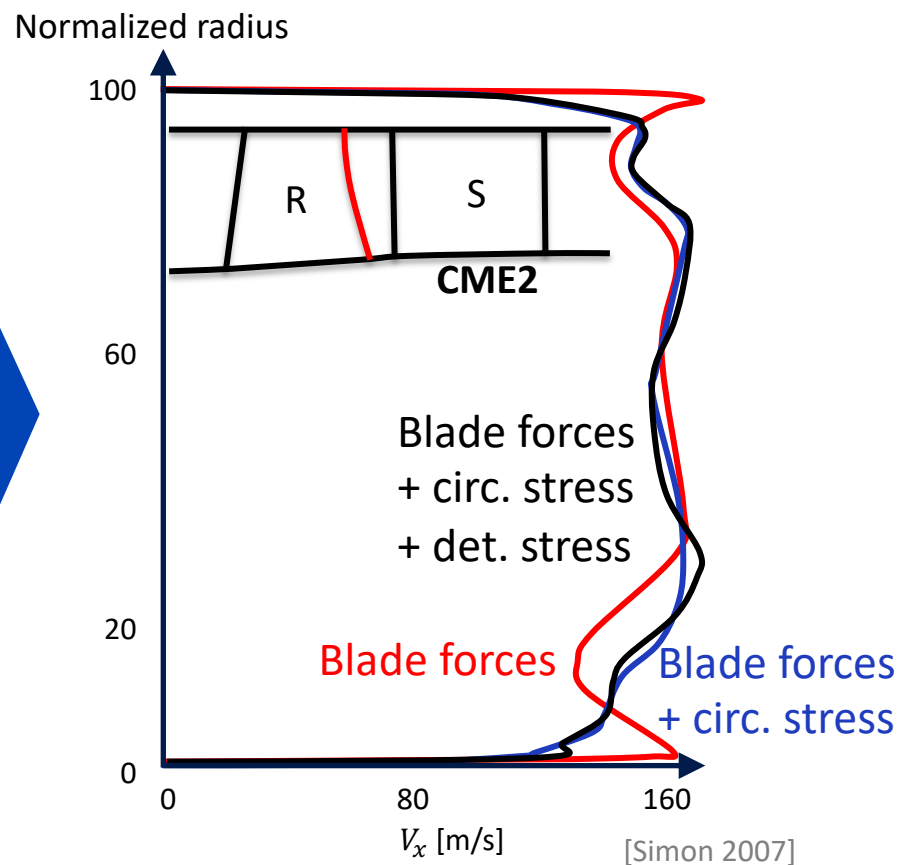
Empiricism/approximation through model

Relative importance of source terms

$$D_t \bar{U}(r, x) = \bar{G}(U, r, x)$$

- Included**
- Reynolds stress
 - Inviscid blade force
 - Viscous blade force
 - Axisymmetric source terms

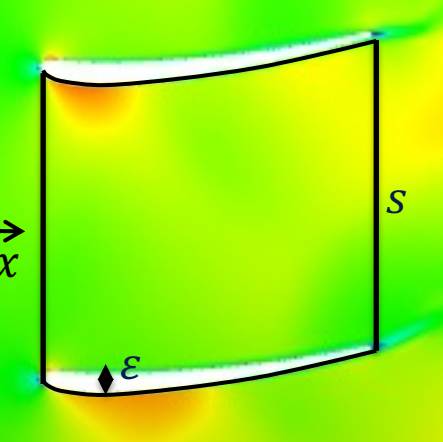
- Reynolds stress
 - Inviscid blade force
 - Viscous blade force
- Major terms**
-
- Circumferential stress
 - Unsteady stress
- Lower importance**
-
- Aperiodic stress
- Generally neglected**



ASTECC: a viscous through-flow model

Circumferential averaged Navier-Stokes equations:

Conservative variables



$$\frac{\partial U}{\partial t} + \frac{1}{b} \frac{\partial \overbrace{b(F - F_v)}^{x\text{-fluxes}}}{\partial x} + \frac{1}{b} \frac{\partial \overbrace{b(G - G_v)}^{r\text{-fluxes}}}{\partial r} = \mathbf{S}$$

Blockage factor

$$b = 1 - \frac{\varepsilon(x)}{s}$$

- Reynolds stress
- Inviscid blade force
- Viscous blade force
- Axisymmetric source terms

Non-intrusive formulation for CFD solver:

$$\frac{\partial U}{\partial t} + \frac{\partial (F - F_v)}{\partial x} + \frac{\partial (G - G_v)}{\partial r} = \mathbf{S} + \frac{(F_v - F)}{b} \frac{\partial b}{\partial x} + \frac{(G_v - G)}{b} \frac{\partial b}{\partial r}$$

Blockage factor terms (known)

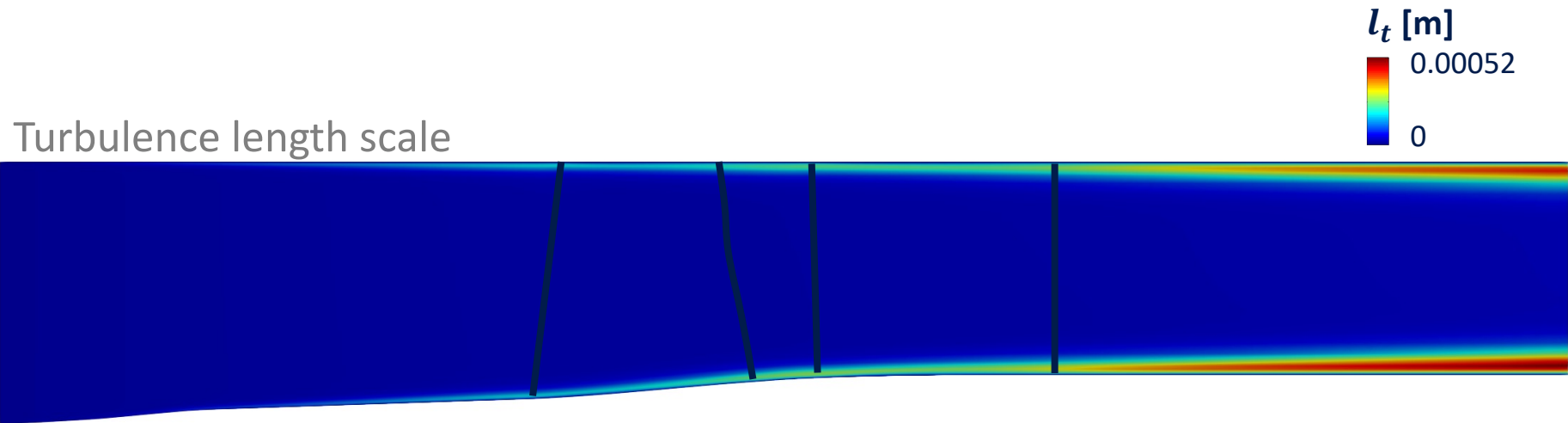
Viscous TF model: closure models

Circumferential averaged Navier-Stokes equations:

$$\frac{\partial \mathbf{U}}{\partial t} + \frac{\partial (\mathbf{F} - \mathbf{F}_v)}{\partial x} + \frac{\partial (\mathbf{G} - \mathbf{G}_v)}{\partial r} = \mathbf{S}$$

- Reynolds stress
- Inviscid blade force
- Viscous blade force
- Axisymmetric source terms

Reynolds stress τ_{reys} : standard turbulence model ($k - l$ Smith)



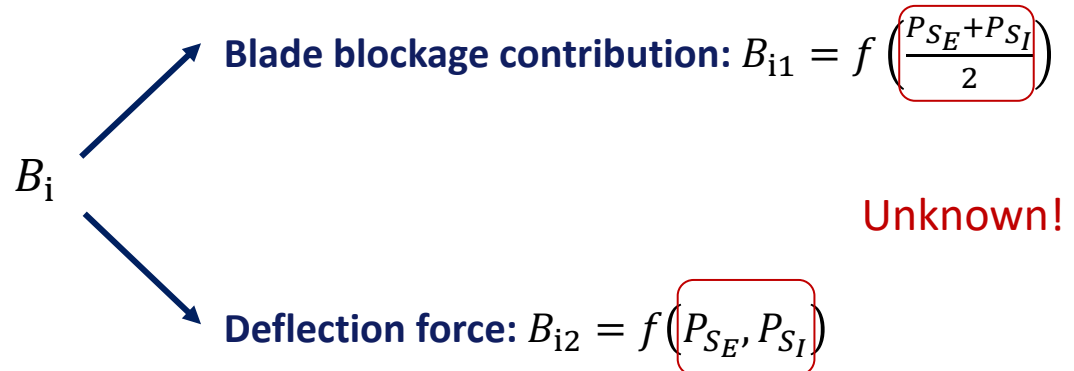
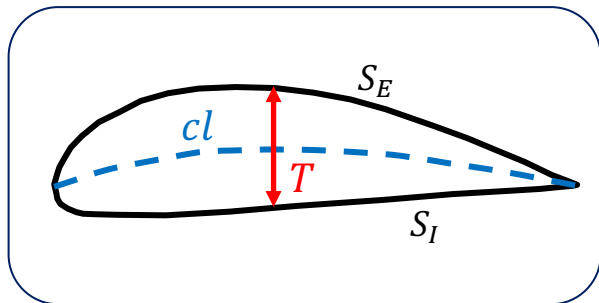
Viscous TF model: closure models

Circumferential averaged Navier-Stokes equations:

$$\frac{\partial \mathbf{U}}{\partial t} + \frac{\partial (\mathbf{F} - \mathbf{F}_v)}{\partial x} + \frac{\partial (\mathbf{G} - \mathbf{G}_v)}{\partial r} = \mathbf{S}$$

- Reynolds stress
- **Inviscid blade force**
- Viscous blade force
- Axisymmetric source terms

Inviscid blade force decomposition B_i :



Closure models: blade forces

$$\frac{\partial U}{\partial t} + \frac{\partial(F - F_v)}{\partial x} + \frac{\partial(G - G_v)}{\partial r} = \mathbf{S}$$

- Reynolds stress
- **Inviscid blade force**
- **Viscous blade force**
- Axisymmetric source terms

Blade blockage B_{i1}

$$B_{i1} = B_{i1} \left(\frac{P_{SE} + P_{SI}}{2} \right)$$



Averaged pressure

$$S_{bi1} = \begin{bmatrix} 0 \\ p \frac{\partial b}{b \partial x} \\ p \frac{\partial b}{b \partial r} \\ 0 \\ 0 \end{bmatrix}$$

Blockage factor: b

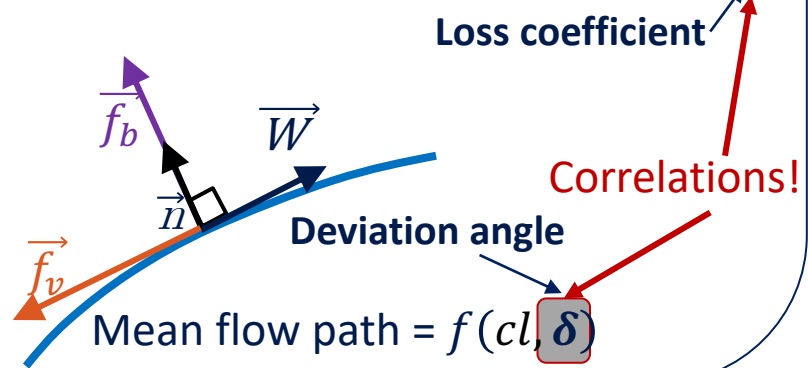
Deflection B_{i2} and viscous force B_v

$$B_{i2} = B_{i2}(P_{SE}, P_{SI}) \quad B_v = B_v(\tau_{SE}, \tau_{SI})$$



Distributed forces:

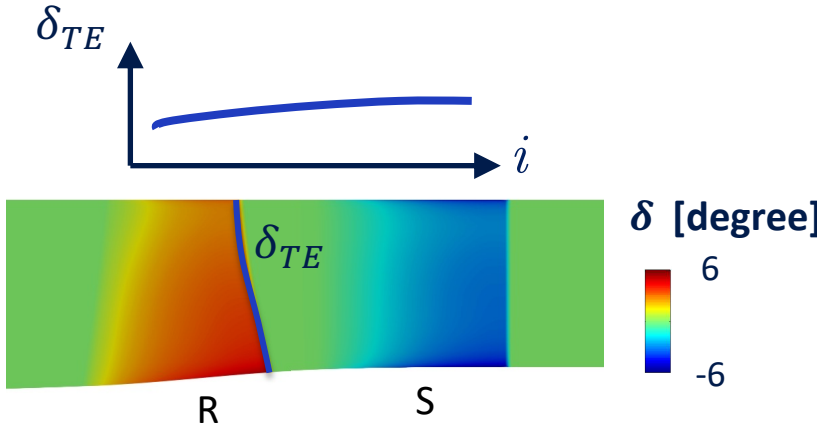
$$\frac{\partial f_b}{\partial \tau} = -C (\vec{W} \cdot \vec{n}) \quad f_v = \rho T \frac{W_m \partial_m s}{W} = f(\omega)$$



Correlations for δ and ω

Deviation angle $\delta \rightarrow$ inviscid blade force

- δ_{TE} from cascade experiments (Lieblein)
- Linear variation with incidence around design conditions
- $\delta = \delta_{TE} \frac{\kappa_{LE} - \kappa}{\kappa_{LE} - \kappa_{TE}}$ ← Blade angle

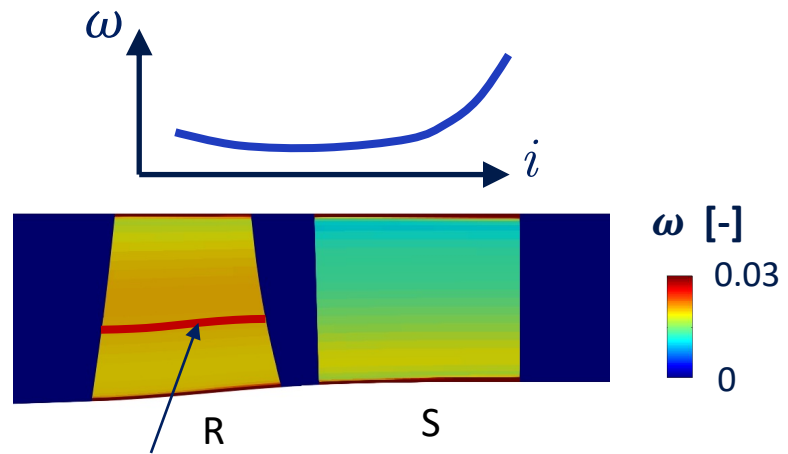


Loss coefficient $\omega \rightarrow$ viscous blade force

- From cascade experiments (Lieblein)



Profile loss only



Constant over streamline (0D)

Outline

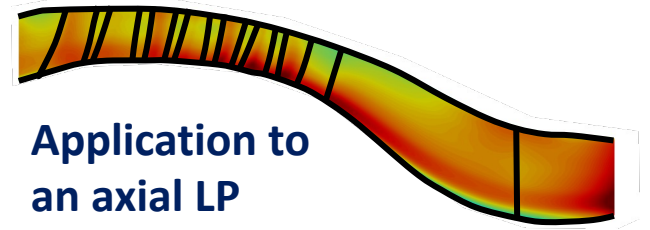
1

Viscous
through-flow model

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2

Model Assessment



Application to
an axial LP
compressor

3

Geometric variability

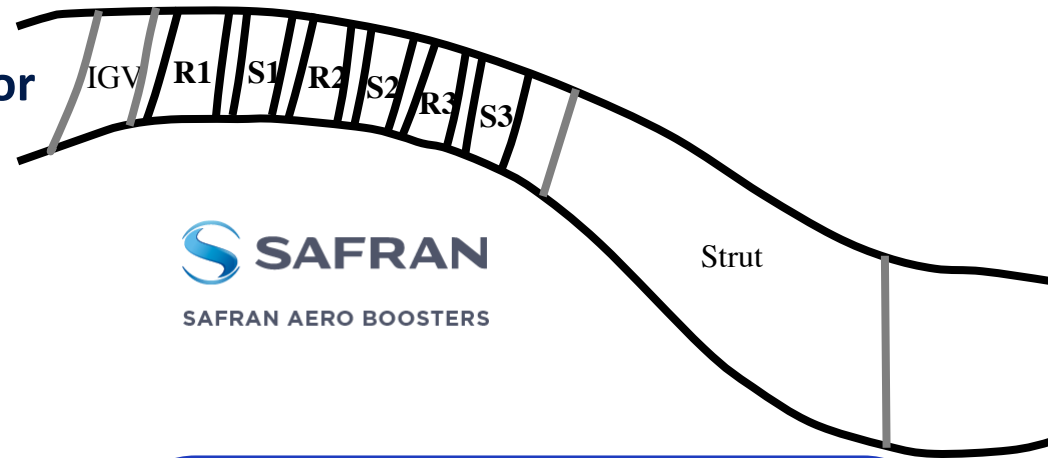


[SAB]

Closure model assessment

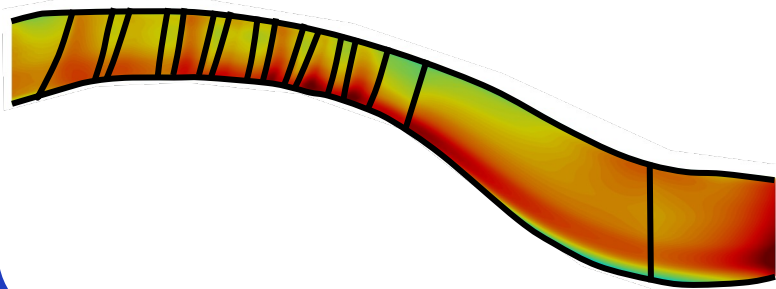
- Quantify closure **model errors**
- Simulations fed with **exact** averaged distribution of δ_{TE}, ω

Test-case : low-pressure compressor

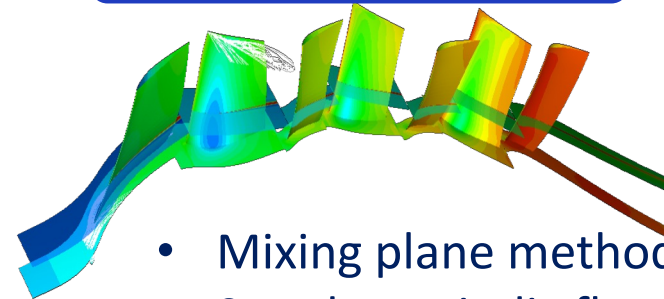


- Highly loaded
- High subsonic Mach number
- 3D modern blades

Through-flow (TF) simulations

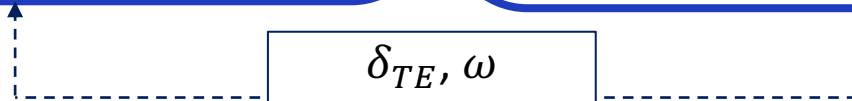


3D RANS simulations

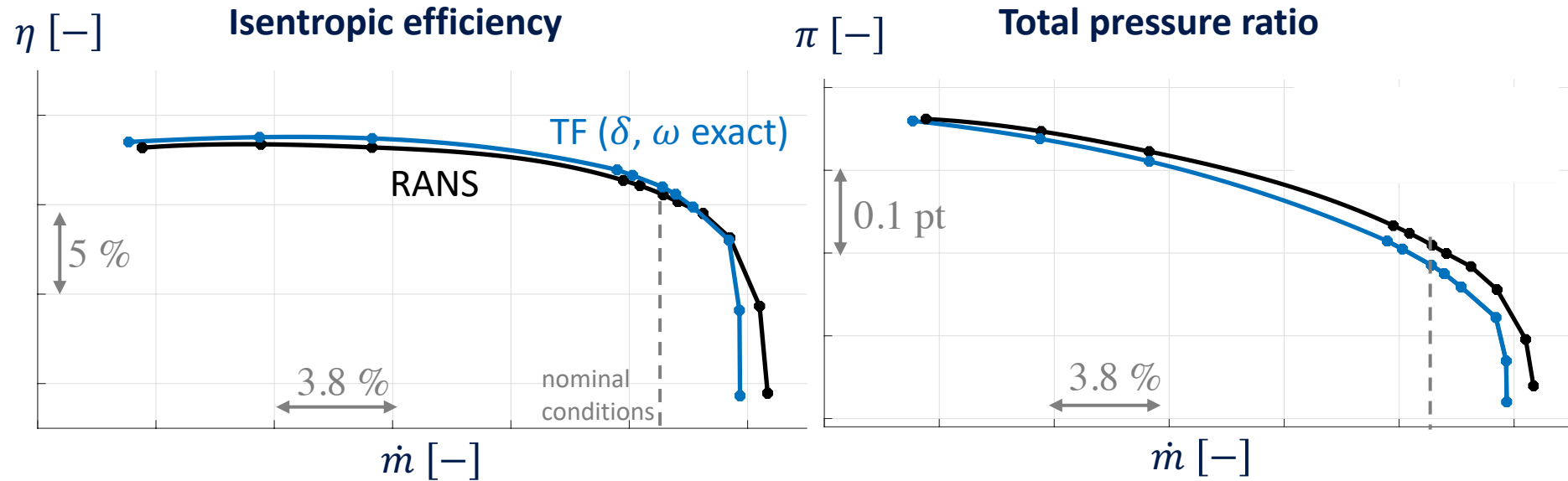


- Mixing plane method
- Steady, periodic flow

δ_{TE}, ω
distributions



Closure model assessment



- Good prediction (low margin)
- 600 times faster
- Sources of errors: τ_{circ} , closure model form, δ_{TE} distribution, blockage assumption, turbulence model...

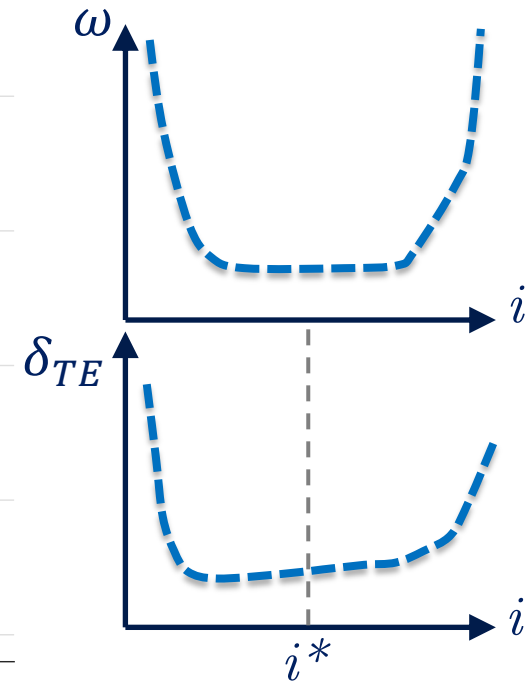
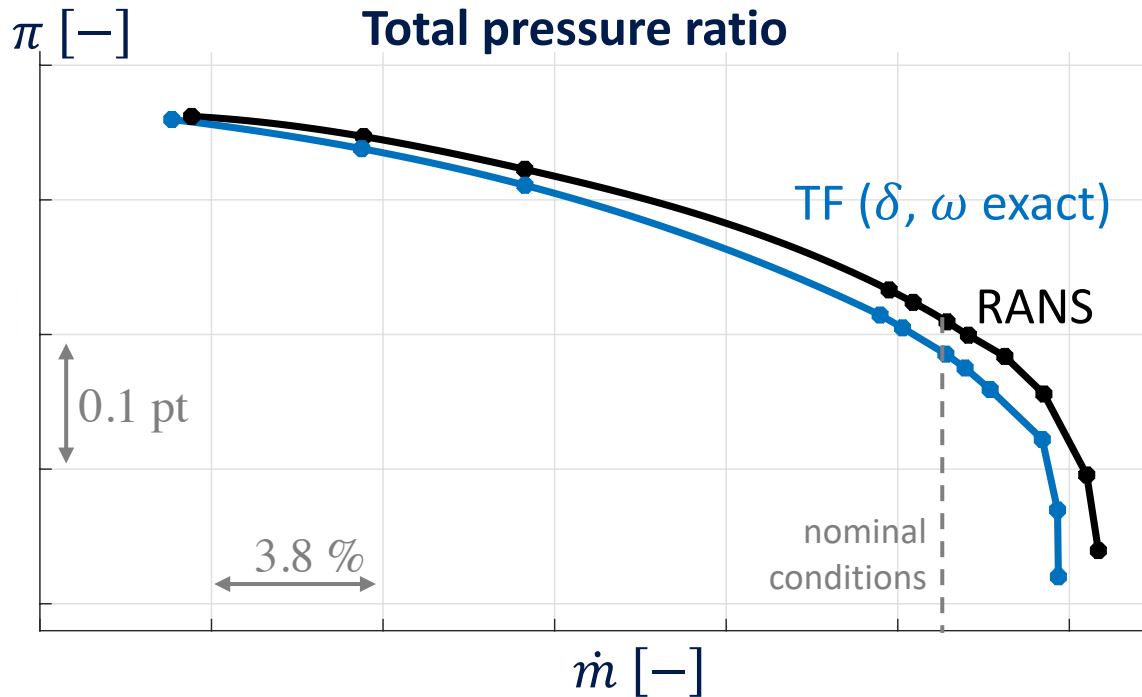


Model able to predict performance

But exact δ, ω unknown in practice...

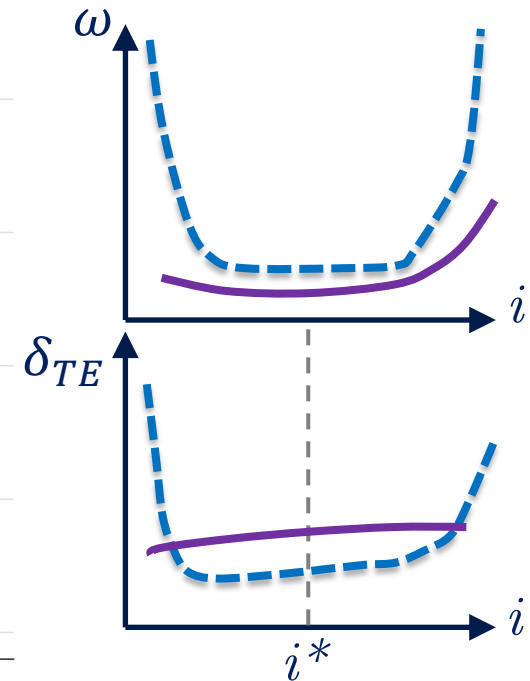
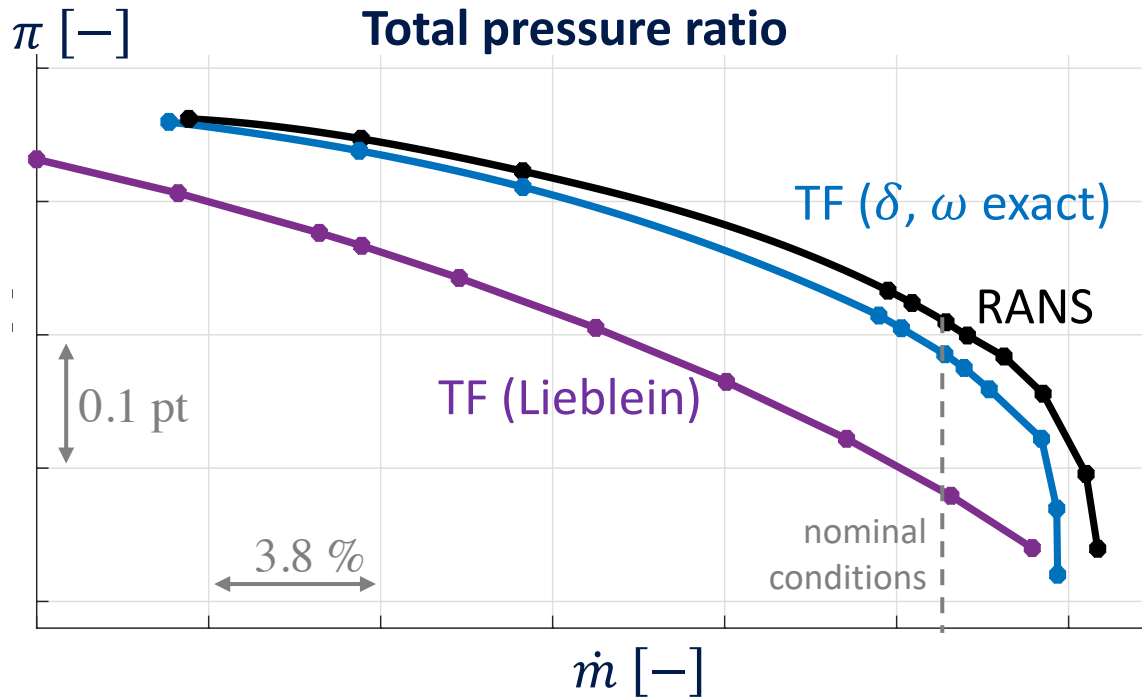
Correlations assessment

- Error quantification of correlations for δ , ω



Correlations assessment

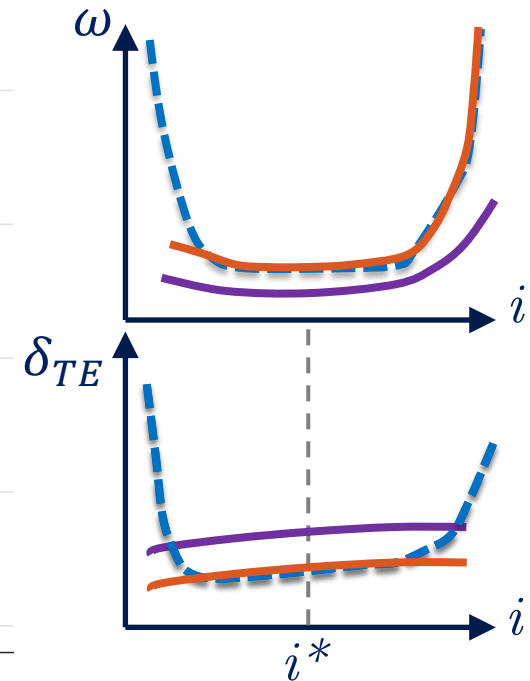
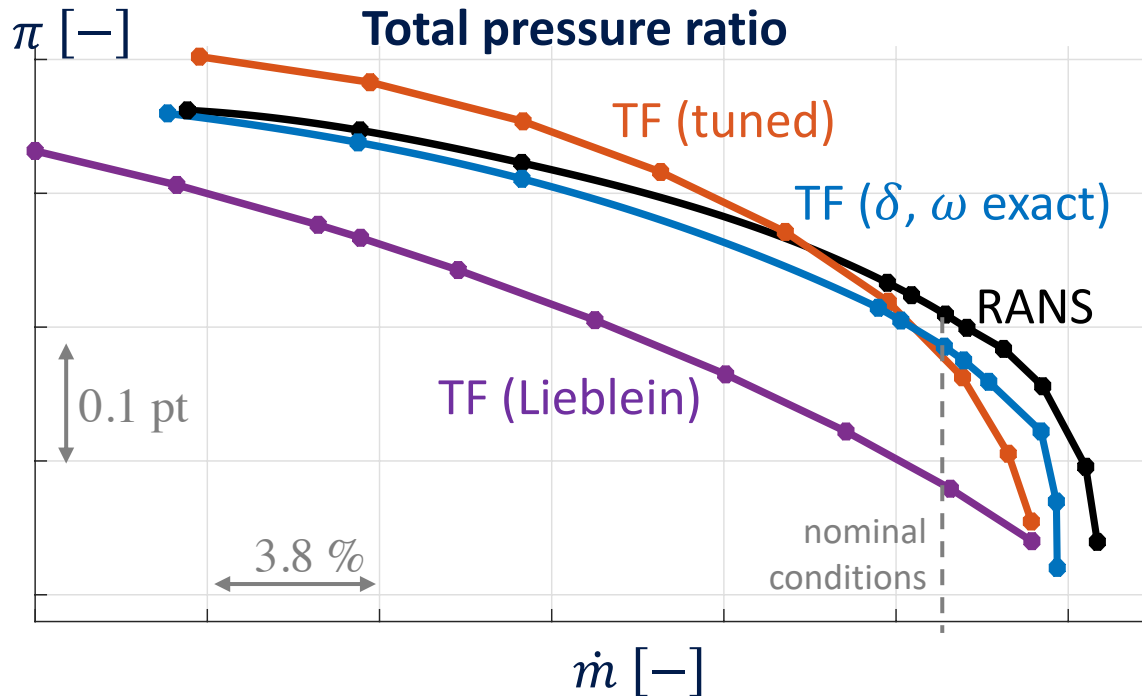
- Error quantification of correlations for δ , ω



- Inaccurate when applied to the modern compressor

Correlations assessment

- Error quantification of correlations for δ , ω



- **Rotor deviation angle correction** → total pressure ratio improvement
- **Mach number effect** added to loss coefficient



Strong dependence of model prediction with respect to correlation accuracy

Outline

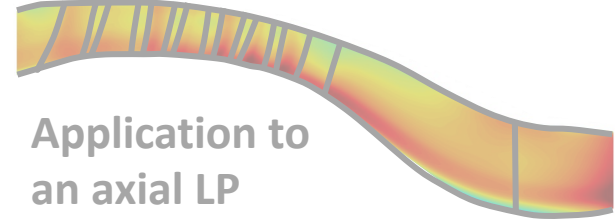
1

Viscous
through-flow model

$$\frac{\partial U}{\partial t} + \frac{\partial(F-F_v)}{\partial x} + \frac{\partial(G-G_v)}{\partial r} = S$$

2

Model Assessment



Application to
an axial LP
compressor

3

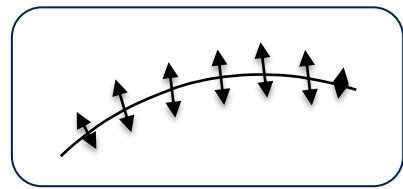
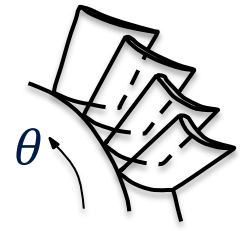
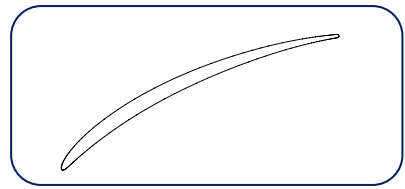
Geometric variability



[SAB]

Geometry in through-flow model

$$D_t U(r, \theta, x, t) = G(U, r, \theta, x, t) \xrightarrow{\theta\text{-averaging}} D_t \bar{U}(r, \theta, x) = \bar{G}(U, r, \theta, x)$$



Camber line coordinates & Thickness distribution

Blade forces

$$b = 1 - \frac{\epsilon}{s}$$

Direct impact

Correlations

$$\delta_{TE}, \omega, i^* = f(\text{geometry, flow quantities})$$

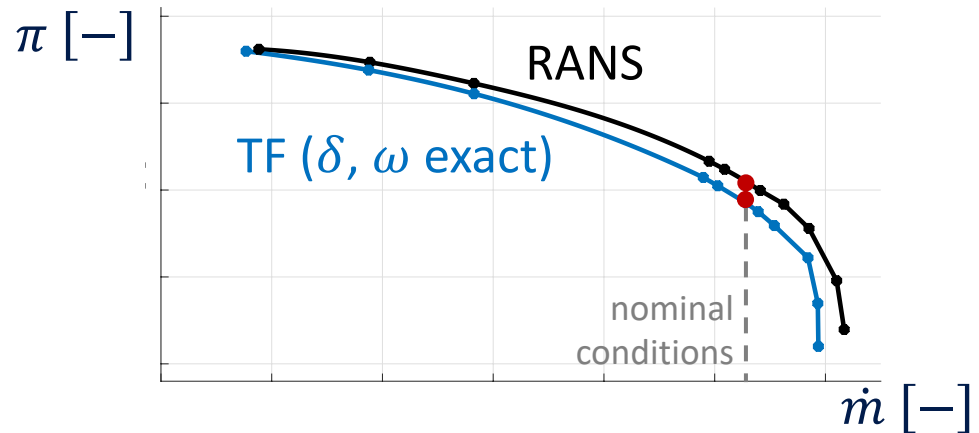
Indirect impact

Geometric variabilities

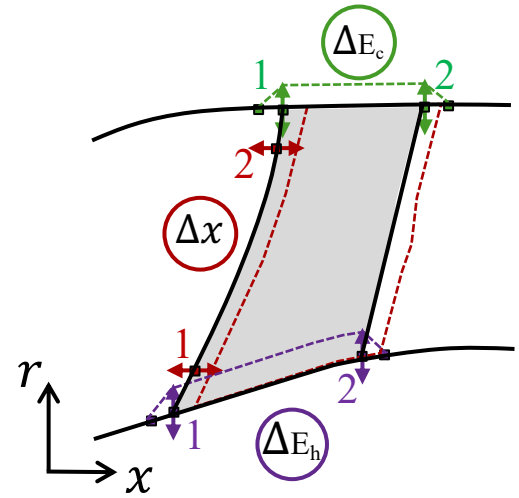
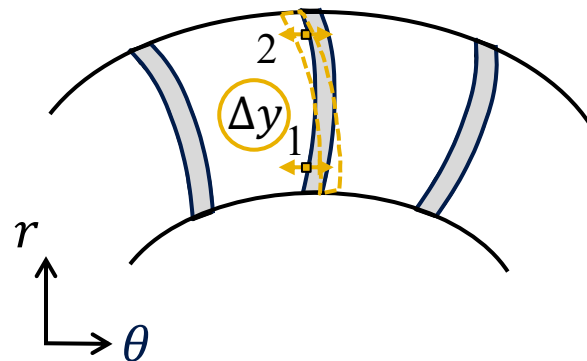
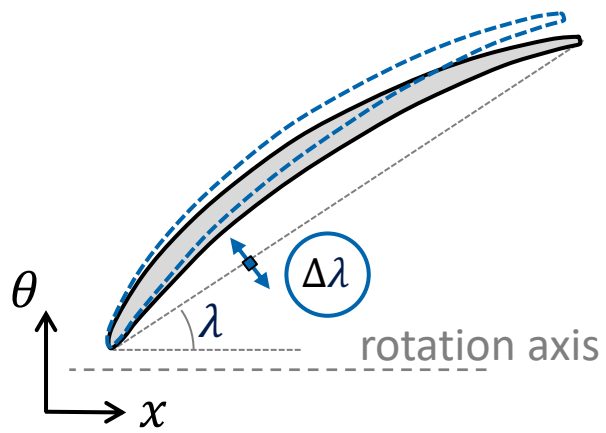
➤ Assess adequacy to predict **performance** variation due to **geometric variabilities**

Preliminary analysis

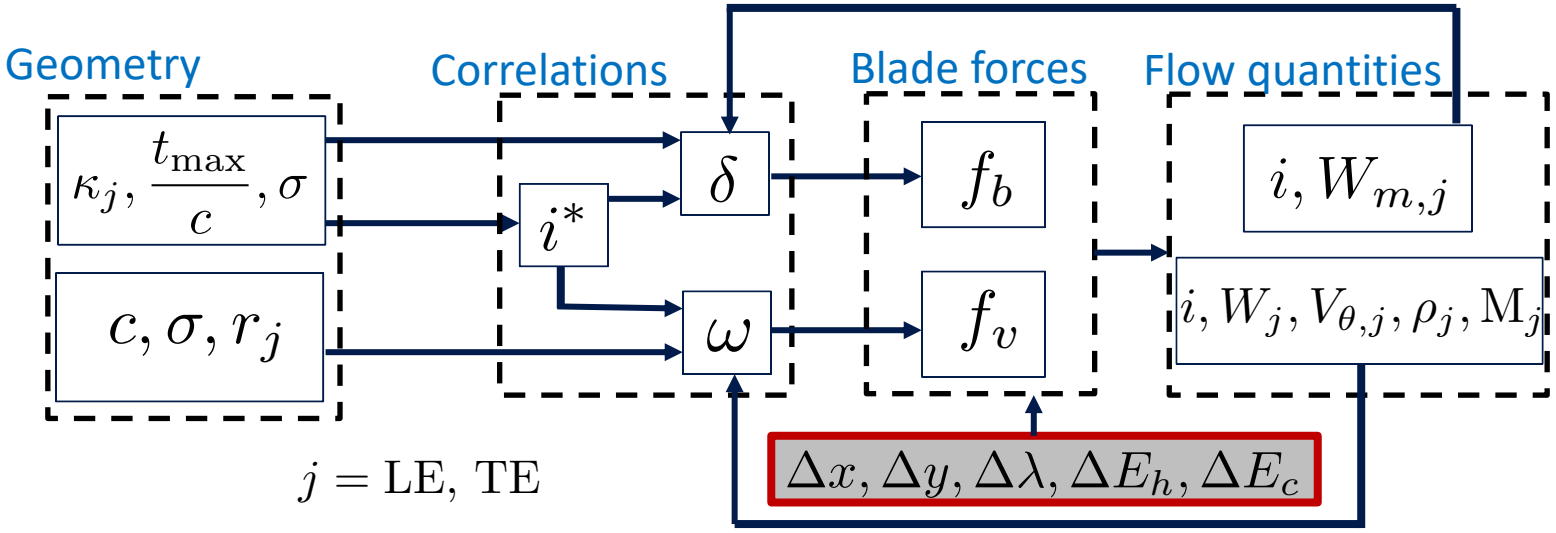
- @ nominal conditions
- Relative variations
- Stator blades



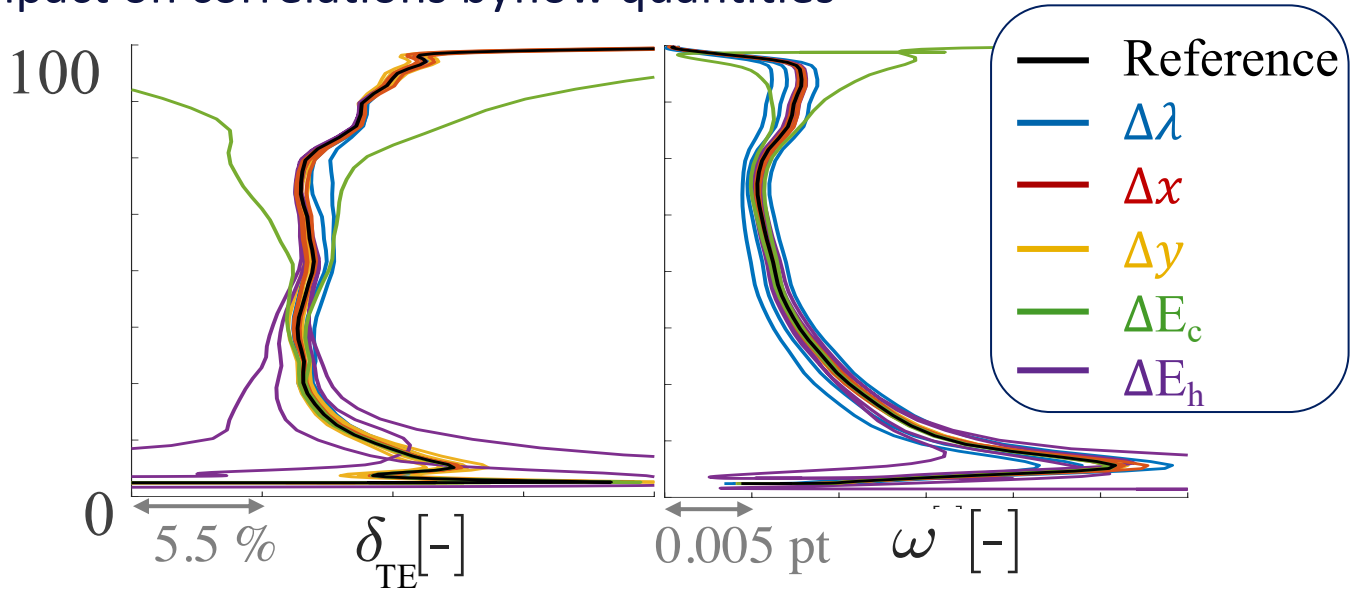
- 3D position of undeformed & endwalls deformation



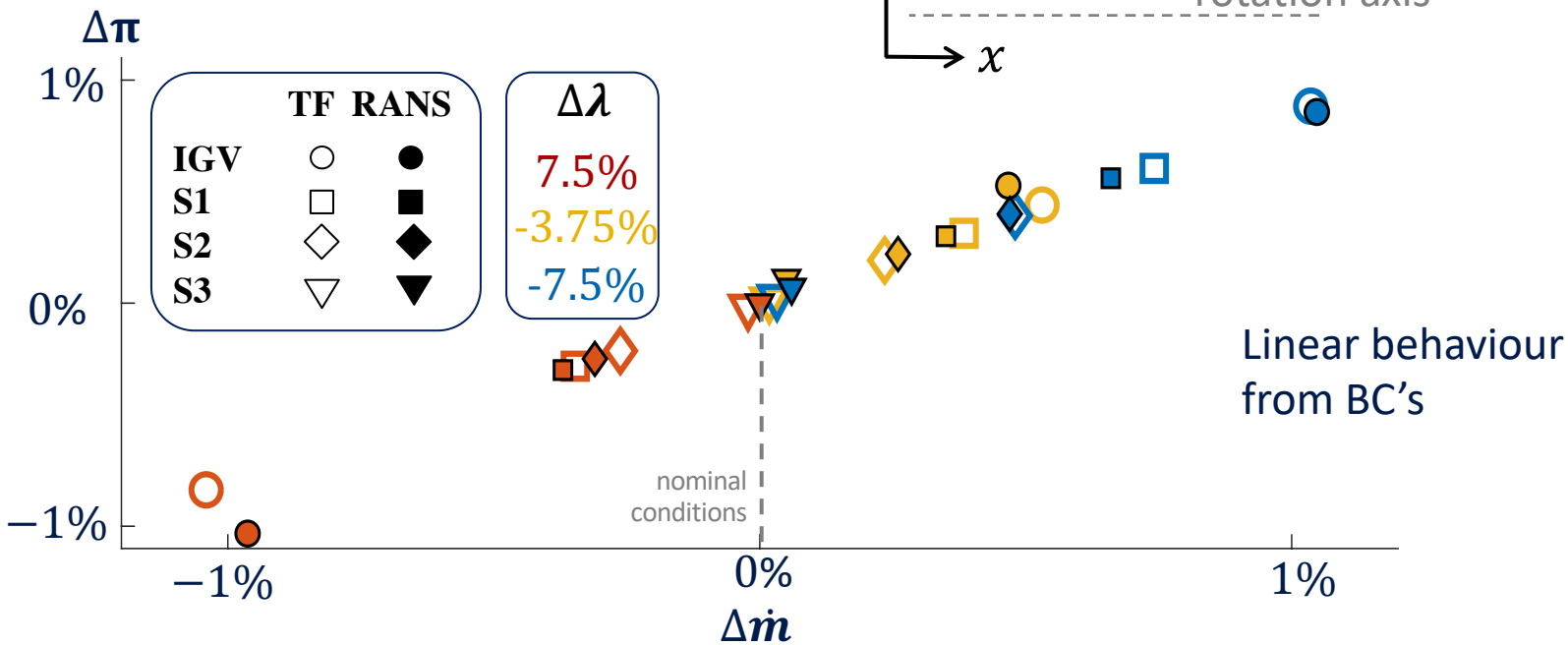
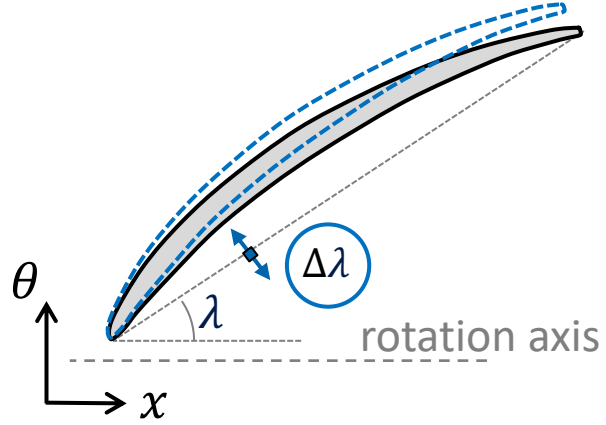
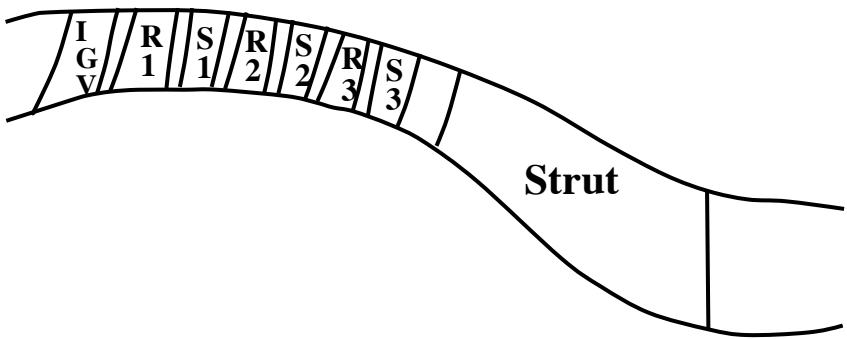
Geometry in through-flow model



➤ Indirect impact on correlations by flow quantities

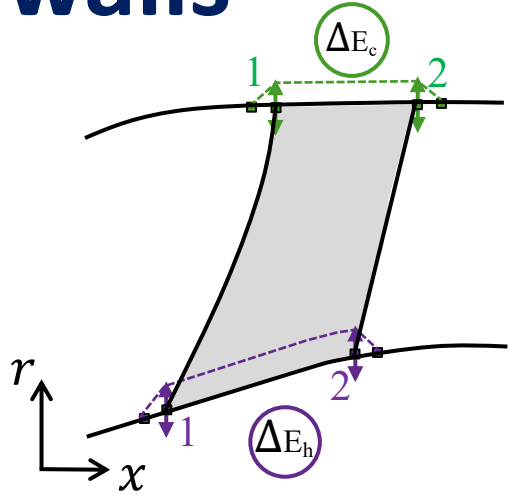
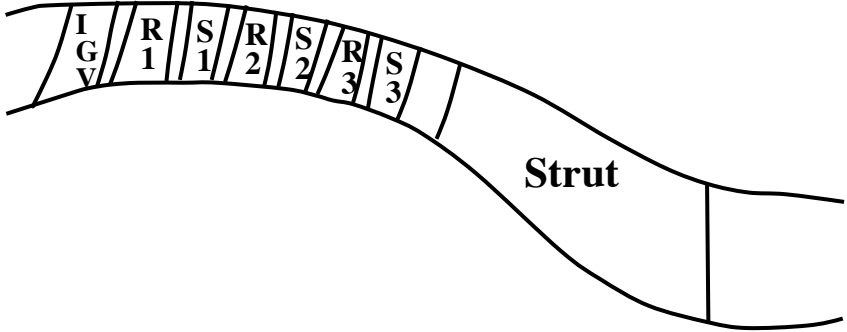


Geometric variabilities: stagger angle

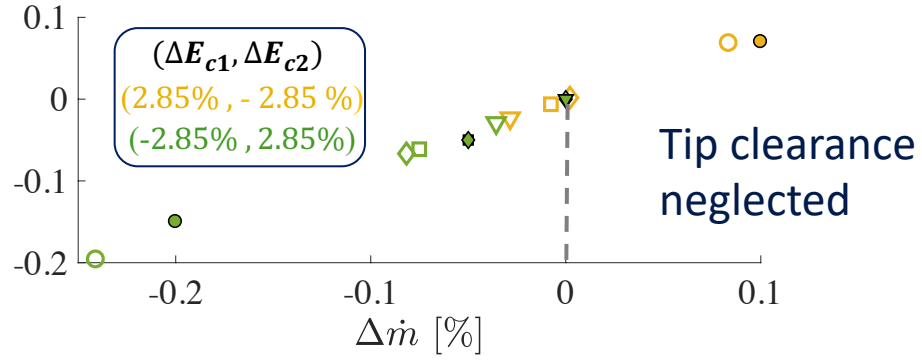
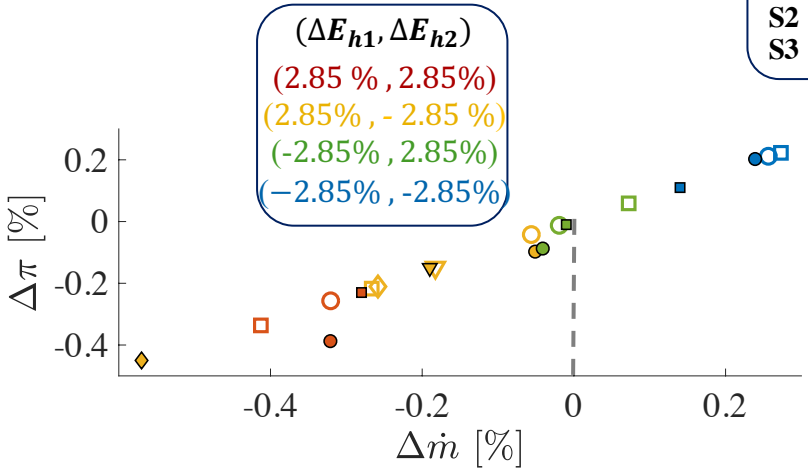


Model able to predict performance variation

Geometric variabilities: endwalls

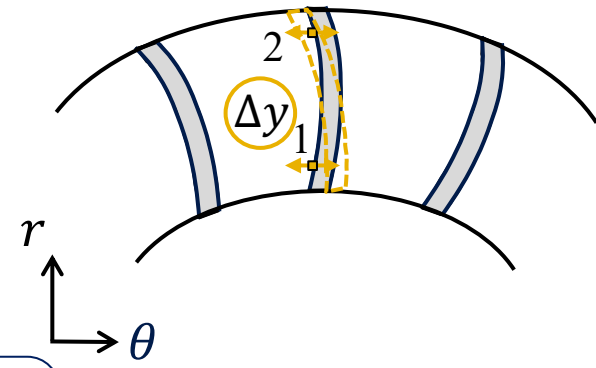
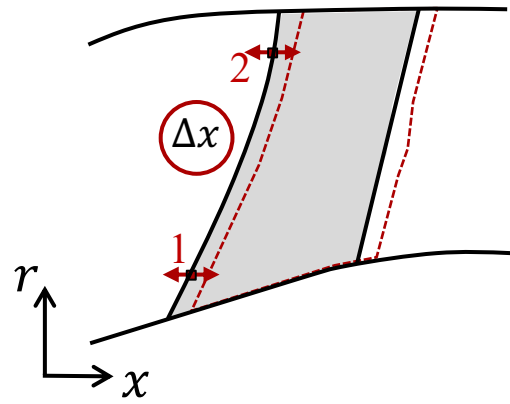


	TF	RANS
IGV	○	●
S1	□	■
S2	◇	◆
S3	▽	▼

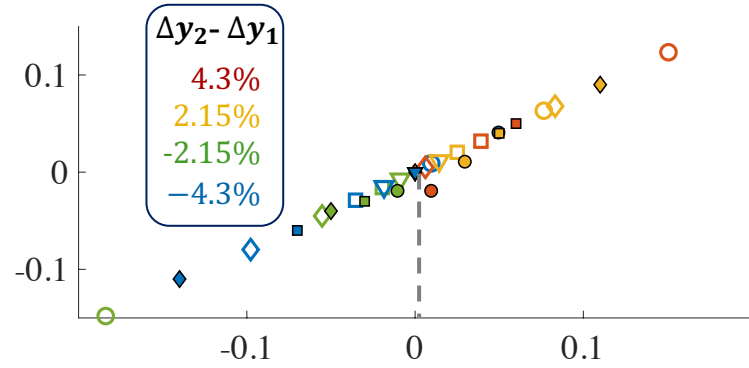
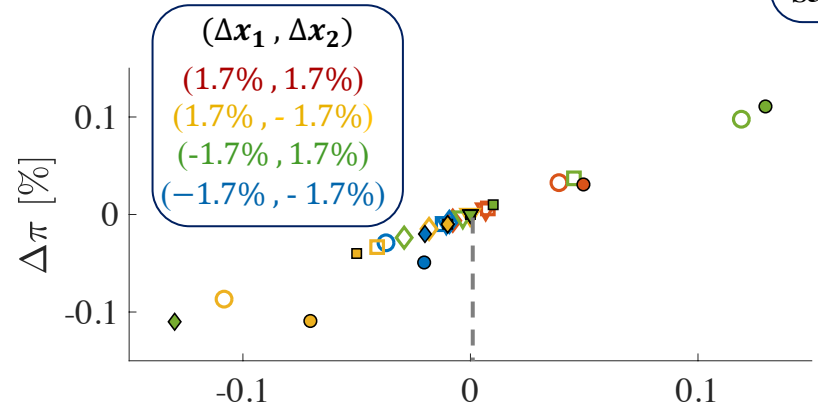


Model able to predict performance variation

Geometric variabilities: blade position



TF RANS	
IGV	○ ●
S1	□ ■
S2	◇ ◆
S3	▽ ▼



Model able to predict performance variation

Conclusion

Through- flow model

- **Reliable low-fidelity** method
- **Good prediction** of performance
- Strong **dependence** between performance prediction and **correlation accuracy**
- Promising approach to drastically **reduce CPU cost** compared to 3D RANS for multi-fidelity approach and UQ

Geometrical variability

- Global good agreement for **performance variation**
- **Promising first step** towards the use of TF modeling for geometric uncertainty quantification

Future work

- Correlation **improvement** @ high incidence
- Thorough analysis of **geometric variability propagation**
- **Strength and weakness** of the model

Acknowledgement

Funding for this research is provided by the Walloon region, under grant no. 7900 in the frame of the project MARIETTA

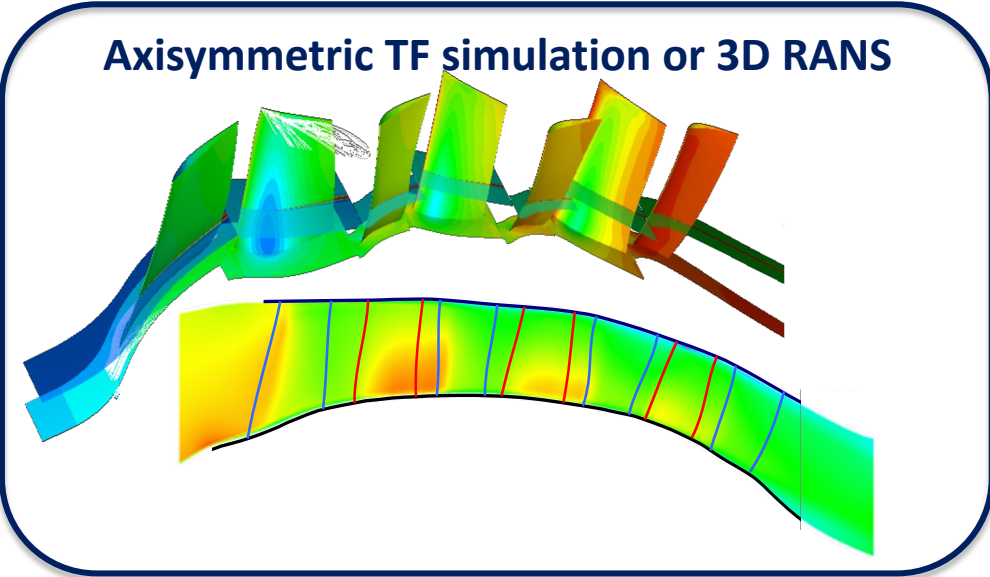


Wallonie

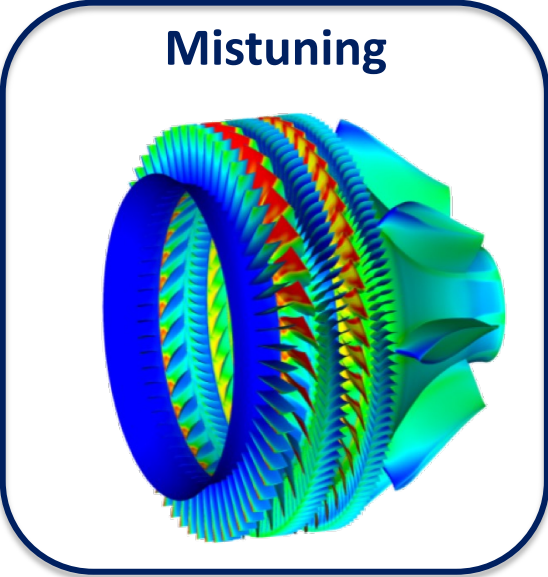


BACK-UP

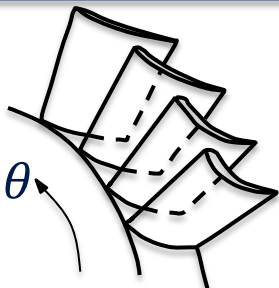
Relevance



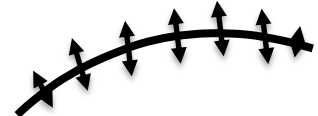
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But

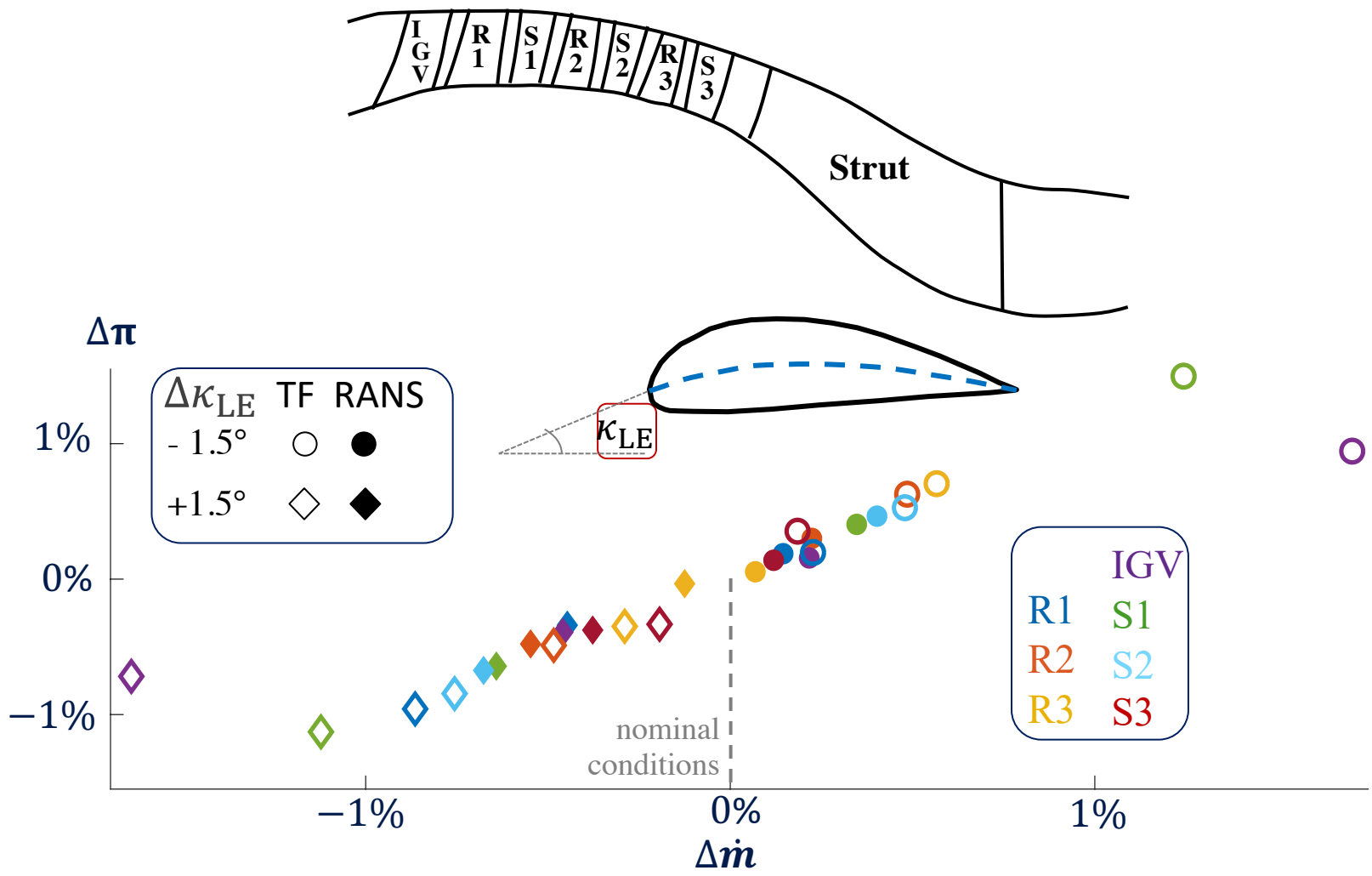


Averaged blade row



- Recurring patterns (tolerances, blade production, ...)
- Low deformation distribution

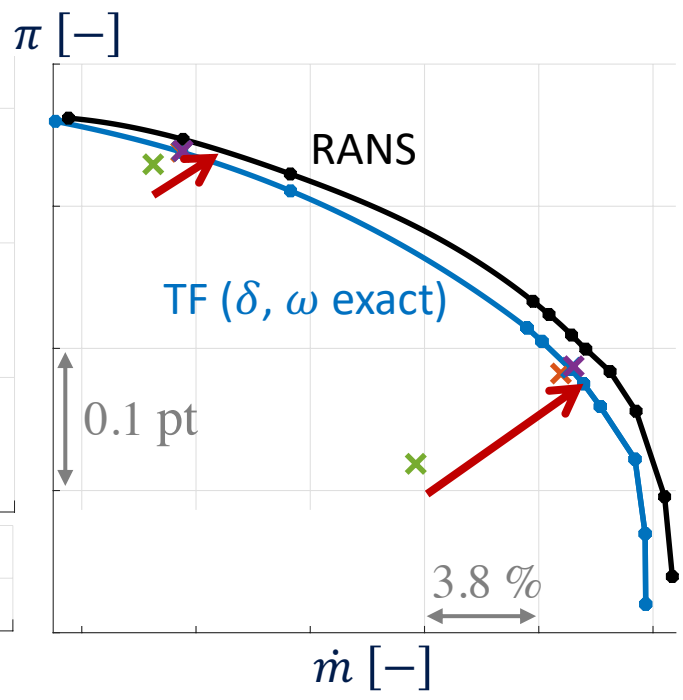
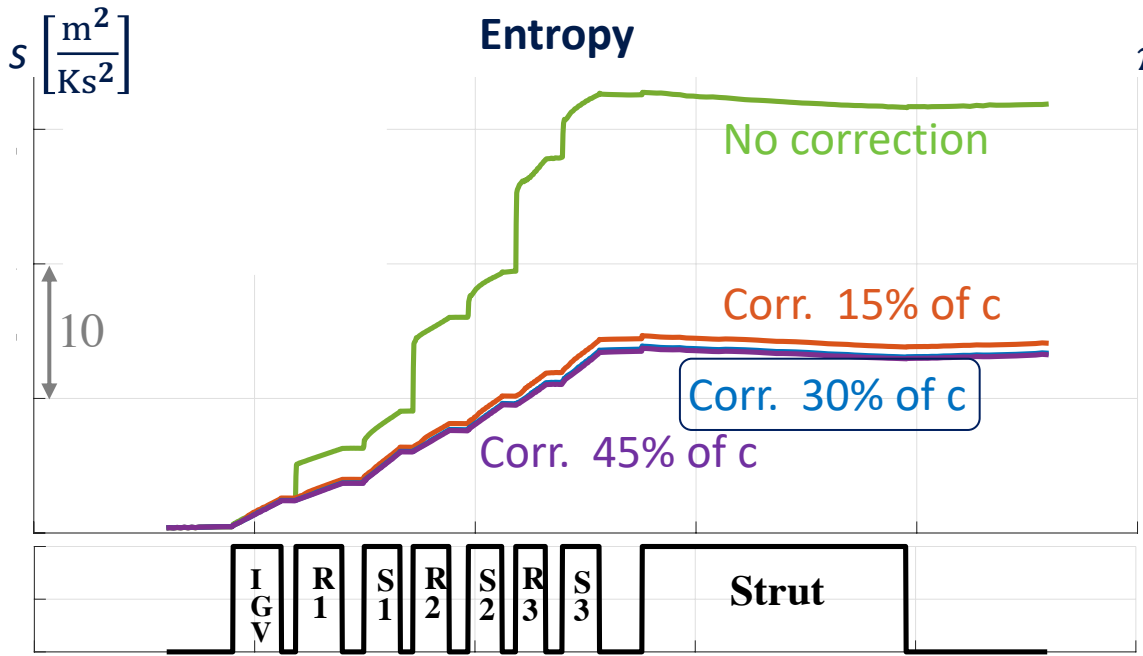
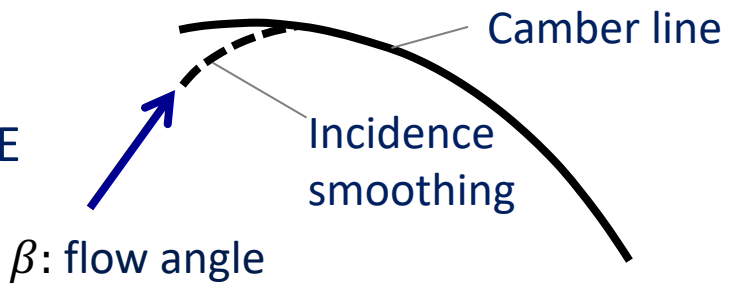
Applications: blade angle variability



Overestimation of performance variability

Incidence correction

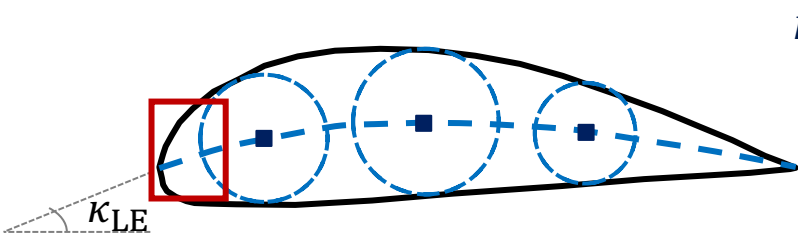
- Avoid flow angle discontinuity
- Modification of blade skeleton @ LE
- Unchanged correlation input



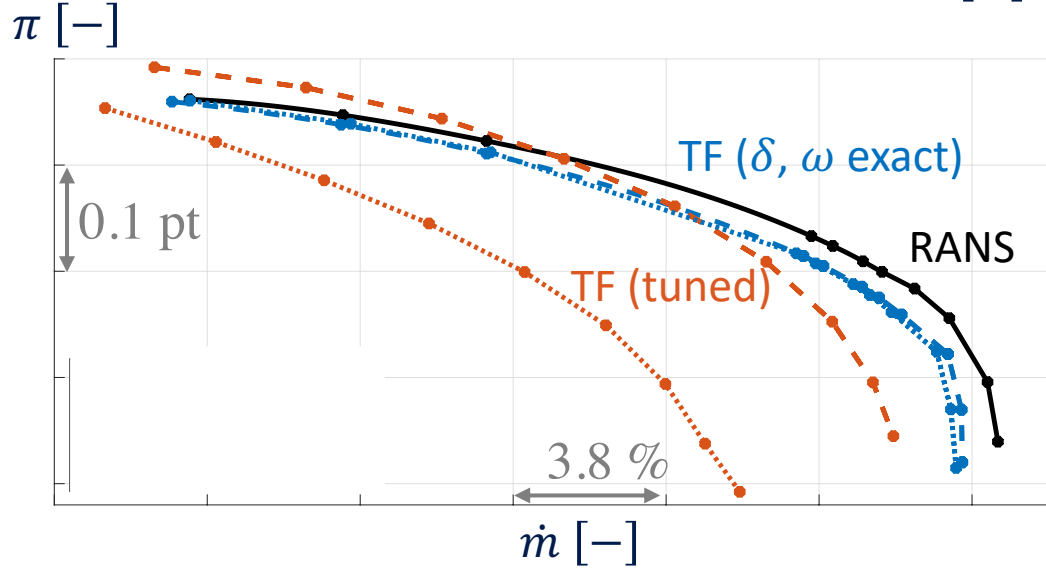
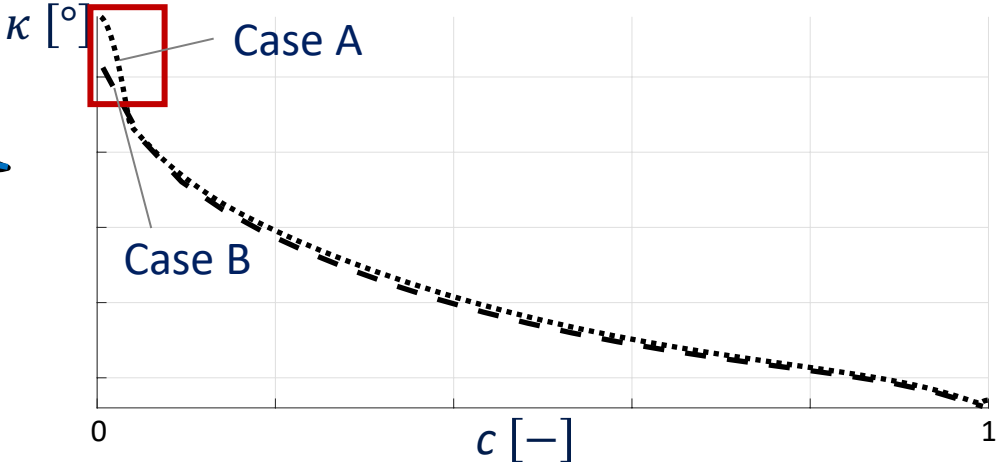
Incidence correction can smooth variability @ LE



Camber line definition



Definition not unique close to LE
 → Large impact on correlation input

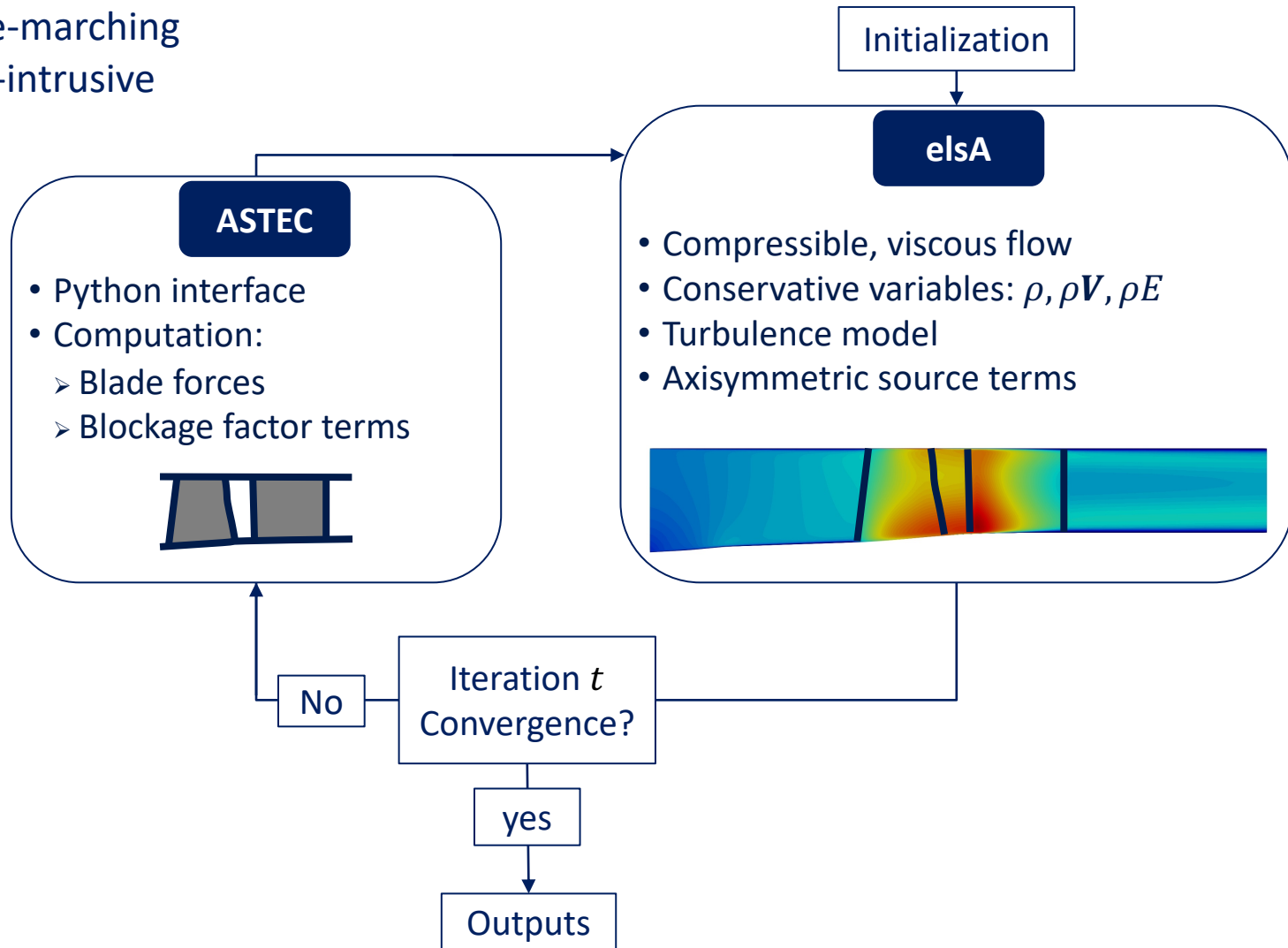


Strong dependence of model prediction with respect to LE blade angle

Viscous through-flow model: ASTEC

Methodology:

- Time-marching
- Non-intrusive



Modern compressor: comparison to RANS

Impact of Mach number

- Minimum-loss incidence angle shifted
- Increase of ω_{\min}
- Narrow range of validity
- Inconsistency between loss validity range and deviation linear range

Correlations **not calibrated** for these flow conditions

Measurements of C4-series cascade

