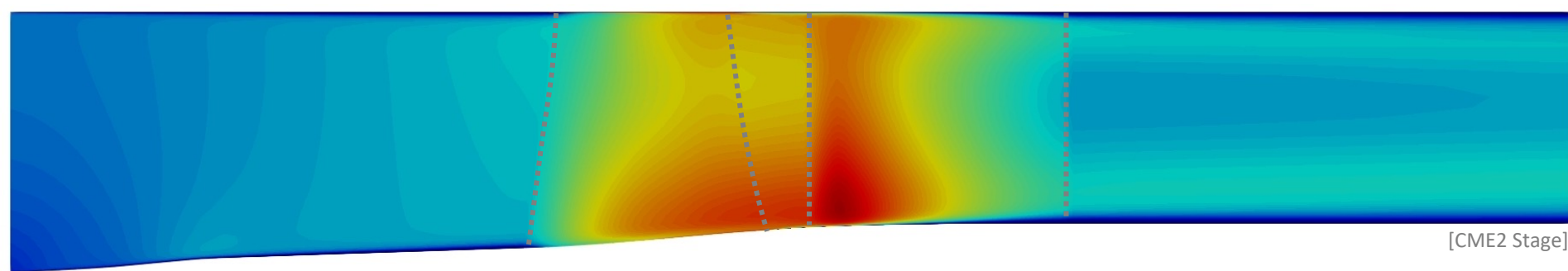


Assessment of geometric variability effects through a viscous through-flow models applied to modern axial-flow compressor blades



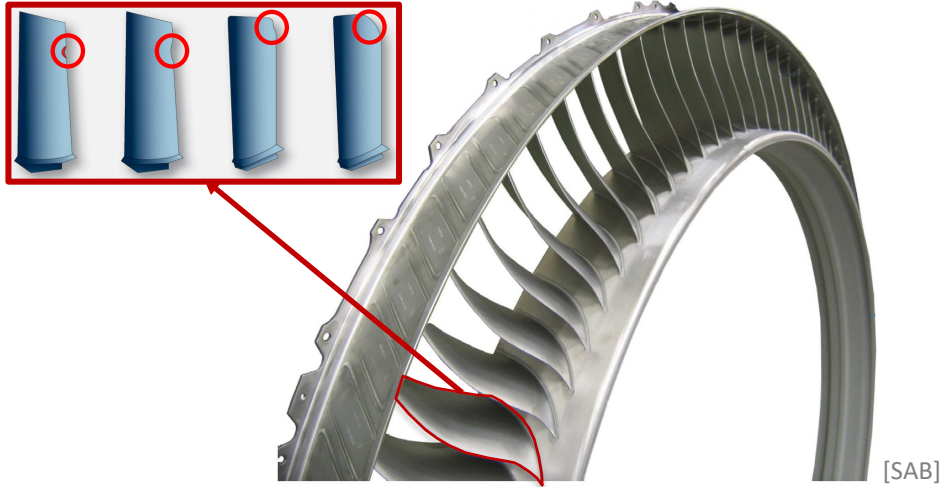
Arnaud Budo⁽¹⁾

Koen Hillewaert⁽¹⁾, Jules Bartholet⁽²⁾,
Maarten Arnst⁽¹⁾, Vincent E. Terrapon⁽¹⁾

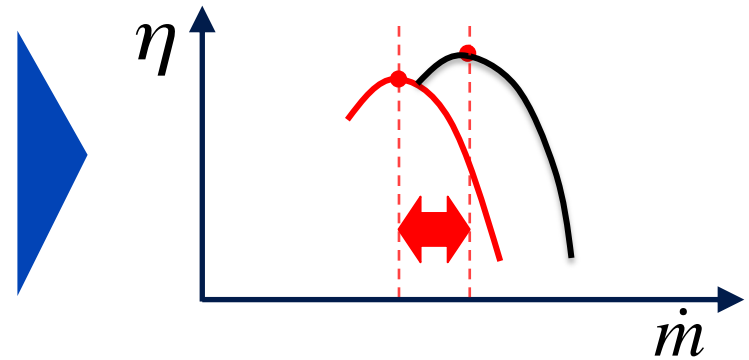
ETC 2023

Context

Geometric variability of low-pressure compressor blades



Performance variation



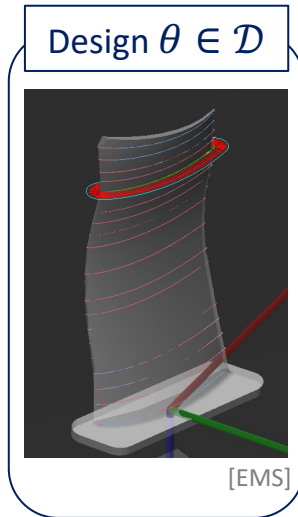
Manufacturing tolerances?

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

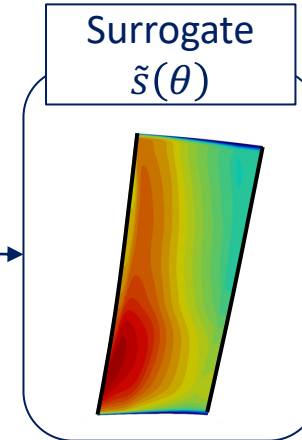
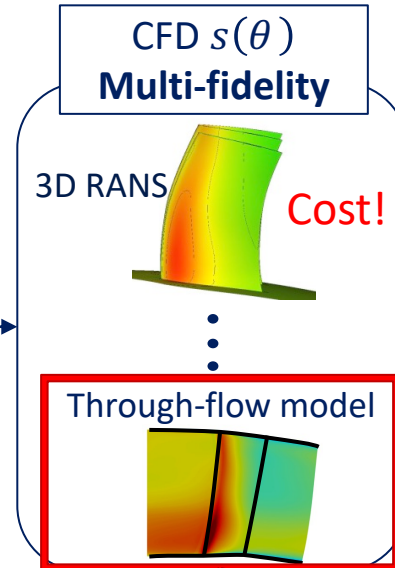
Trade-off
Cost $\Leftrightarrow$ performance

Methodology & objectives

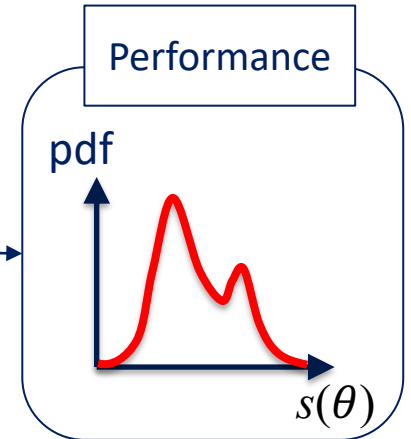
Characterization



Propagation



Qualification



Able to predict performance?

Through-flow model validation

- Low-fidelity approach
- Choice of model correlations

Able to capture variability effects?

Geometrical variability

- Sensitivity analysis
- (Uncertainty quantification)

Outline

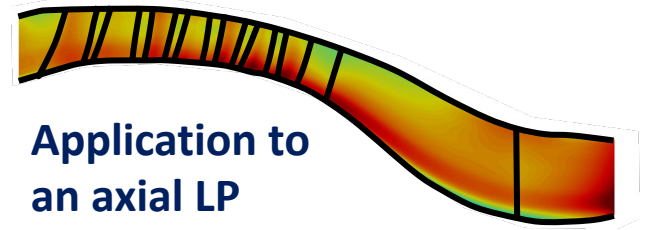
1

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

Geometrical variability



[SAB]

Outline

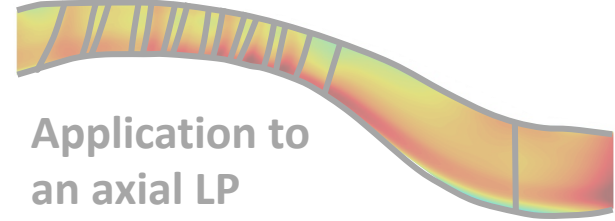
1

Viscous through-flow model

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

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Model Assessment



Application to
an axial LP
compressor

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Geometrical variability



[SAB]

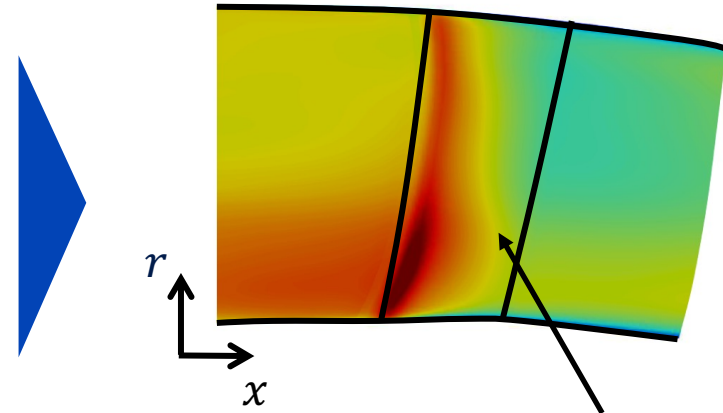
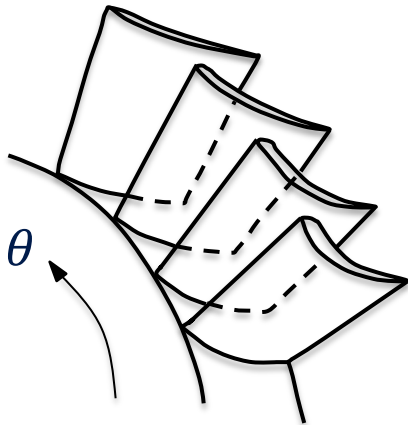
Through-flow model

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

$\xrightarrow{\theta\text{-averaging}}$

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

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

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

Governing equations

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

Closure models

- Exhaustive?
- Assumptions?
- Correlations?

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

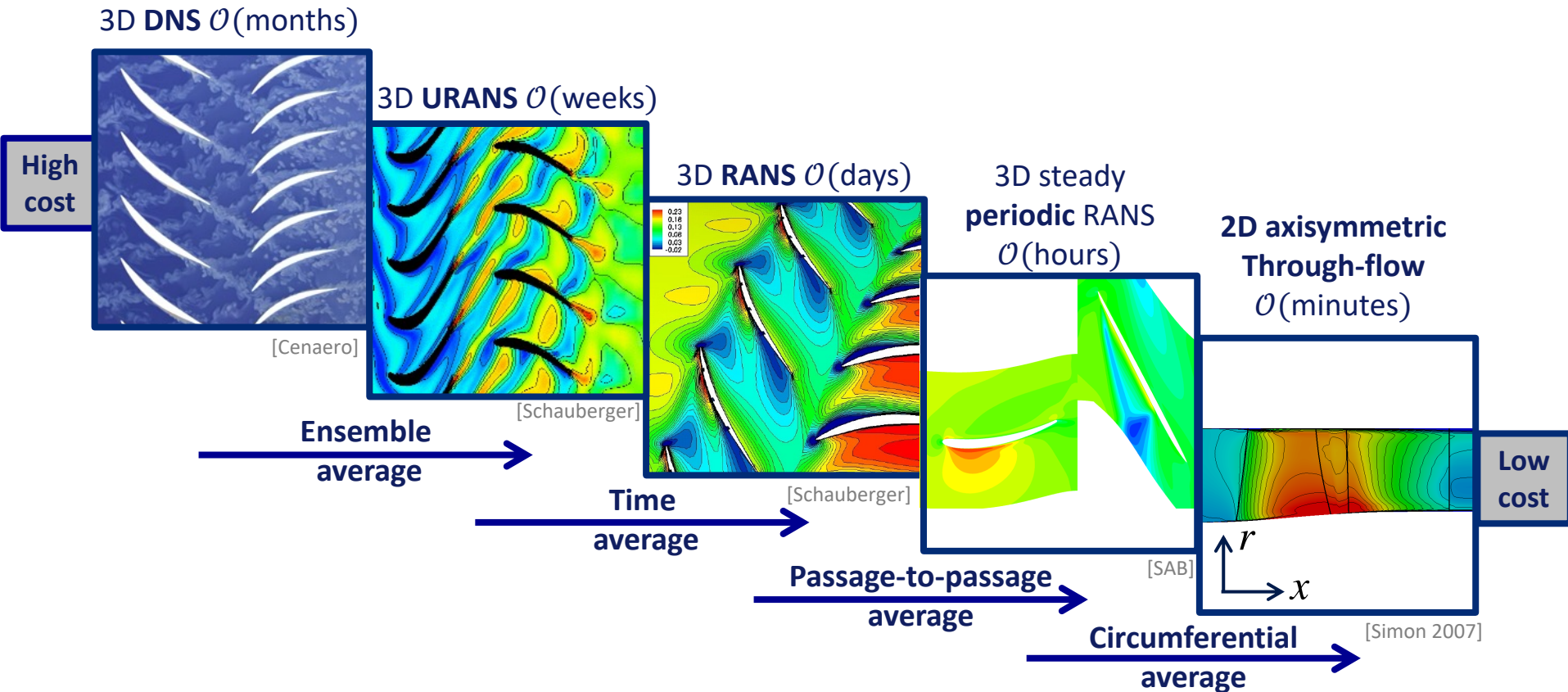
→ **characterize error in a rigorous and exhaustive way**

Adamczyk's cascade

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

- **Exact mathematical** formulation of source terms
- **Robust** and **exhaustive** definition of closures
- **NSE-based equations**

Unclosed!



Adamczyk's cascade: unclosed terms

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

Unclosed terms:

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

**Non-linear
equations**

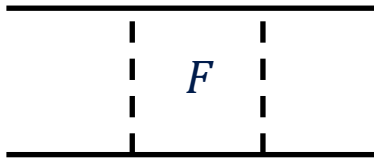
τ_{reys} ➤ Reynolds

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**

Normalized radius

100

60

20

0

0

80

V_x [m/s]

[Simon 2007]

Blade forces

Blade forces
+ circ. stress

Blade forces
+ circ. stress
+ det. stress

CME2

R

S

Viscous TF model: closure models

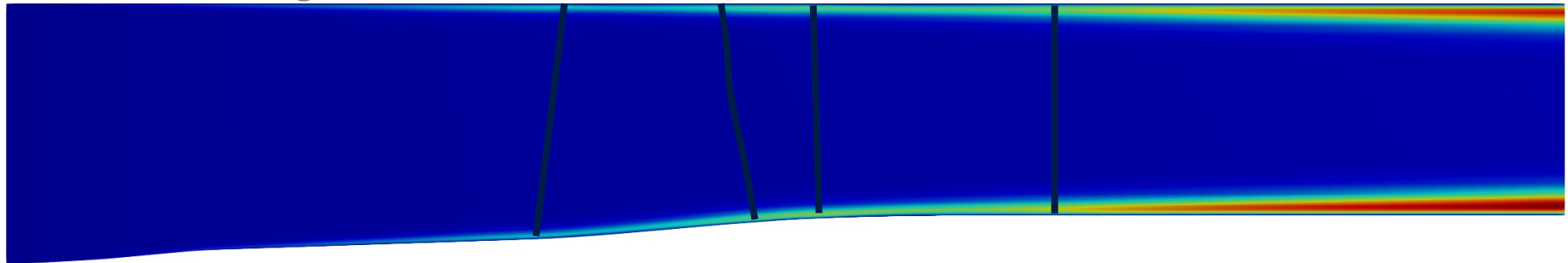
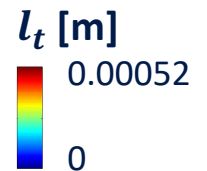
Circumferential averaged Navier-Stokes equations:

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

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

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

Turbulence length scale



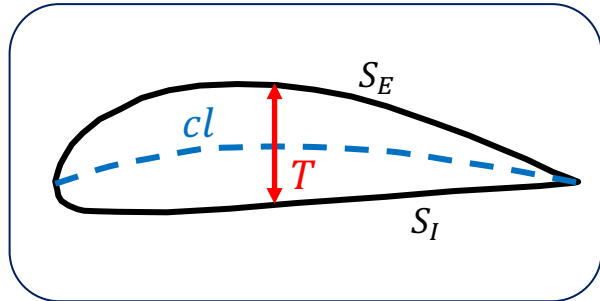
Viscous TF model: closure models

Circumferential averaged Navier-Stokes equations:

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

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

Inviscid blade force decomposition B_i :



B_i

Blade blockage contribution: $B_{i1} = f\left(\frac{P_{S_E} + P_{S_I}}{2}\right)$

Deflection force: $B_{i2} = f(P_{S_E}, P_{S_I})$

Unknown!

Closure models: blade forces

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

- 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

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

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:

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

Loss coefficient

Correlations!

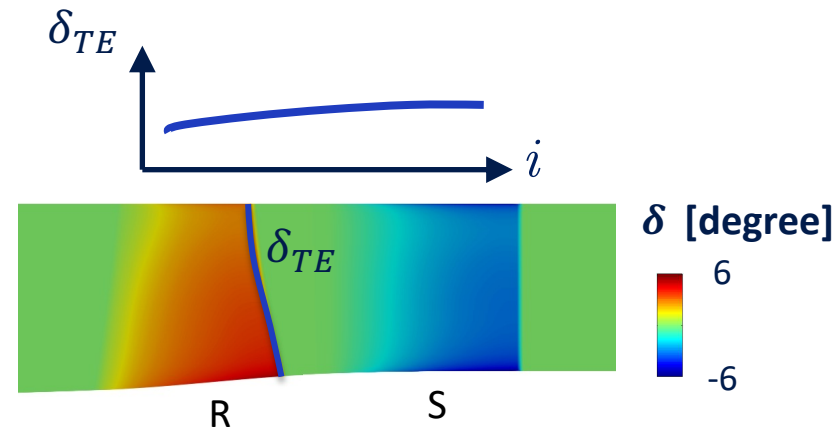
Deviation angle

Mean flow path = $f(cl, \delta)$

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

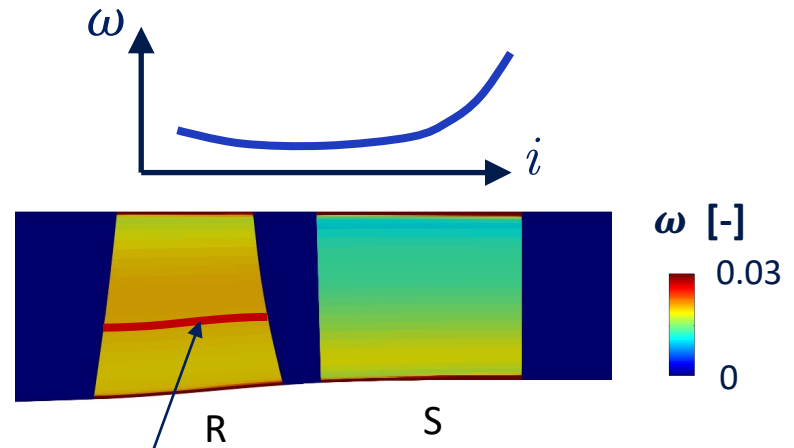


Loss coefficient $\omega \rightarrow$ viscous blade force

- From cascade experiments (Lieblein)



Profile loss only



Constant over streamline (0D)

Outline

1

Viscous
through-flow model

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

2

Model Assessment



Application to
an axial LP
compressor

3

Geometrical variability



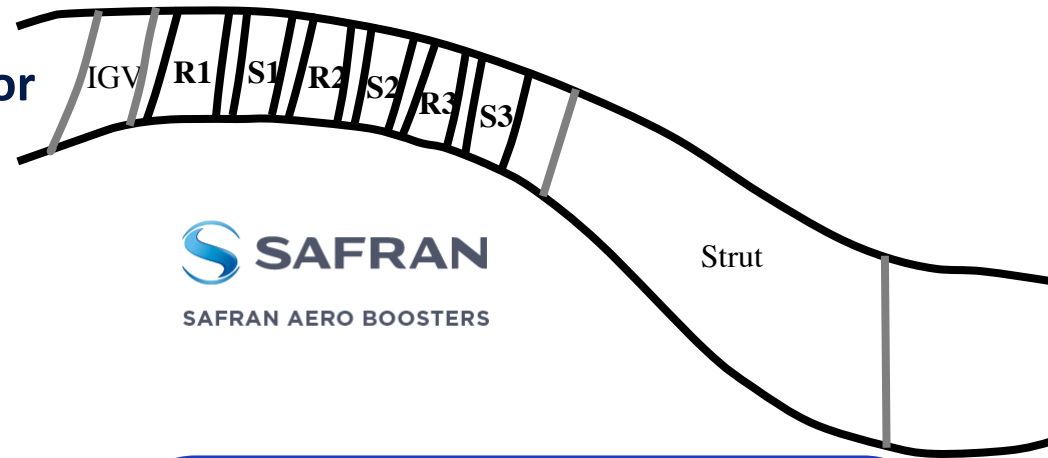
[SAB]

Closure model assessment

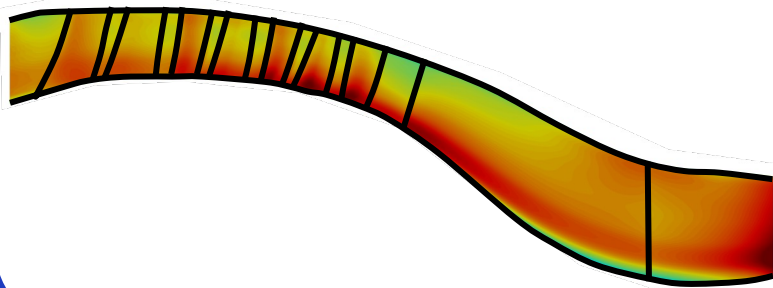
- Quantify closure **model errors**
- Simulations fed with **exact** distribution of δ_{TE} , ω

Test-case : low-pressure compressor

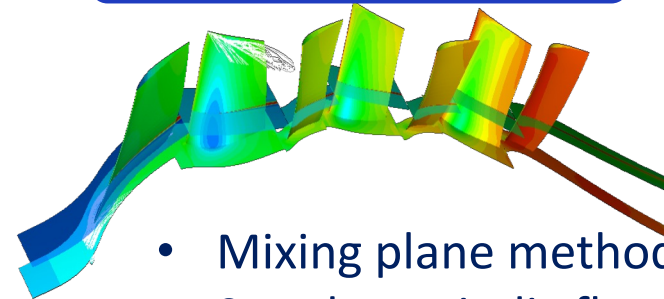
- Highly loaded
- High subsonic Mach number
- 3D modern blades



**Through-flow (TF)
simulations**

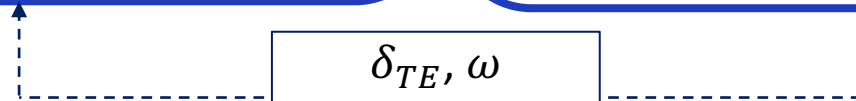


3D RANS simulations

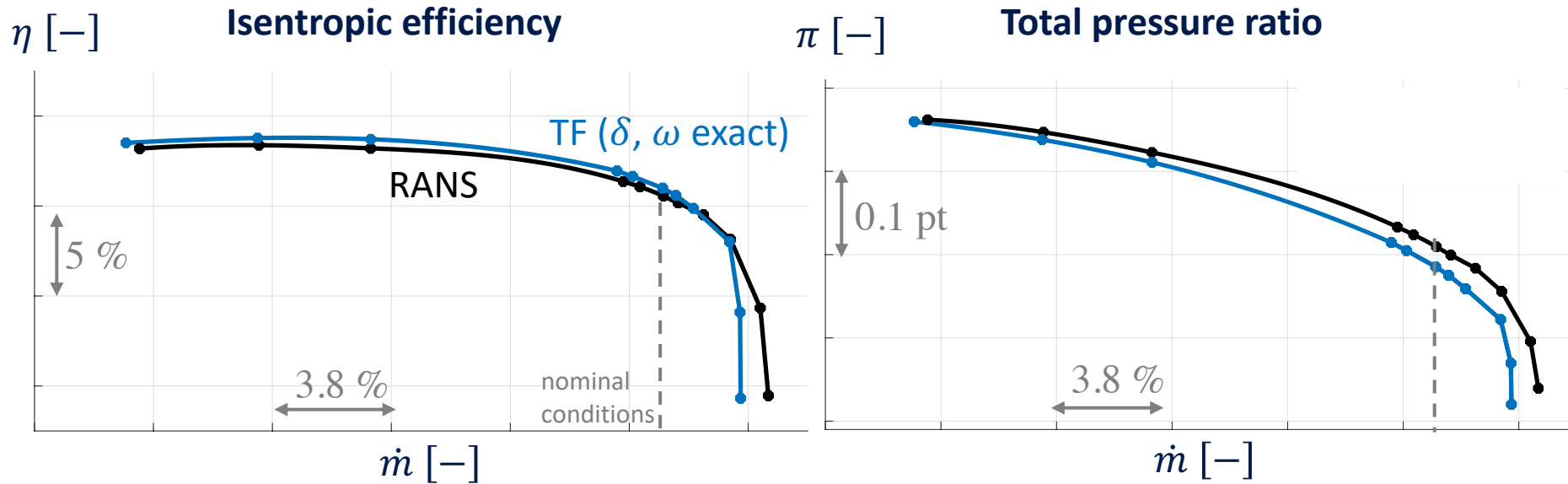


- Mixing plane method
- Steady, periodic flow

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...

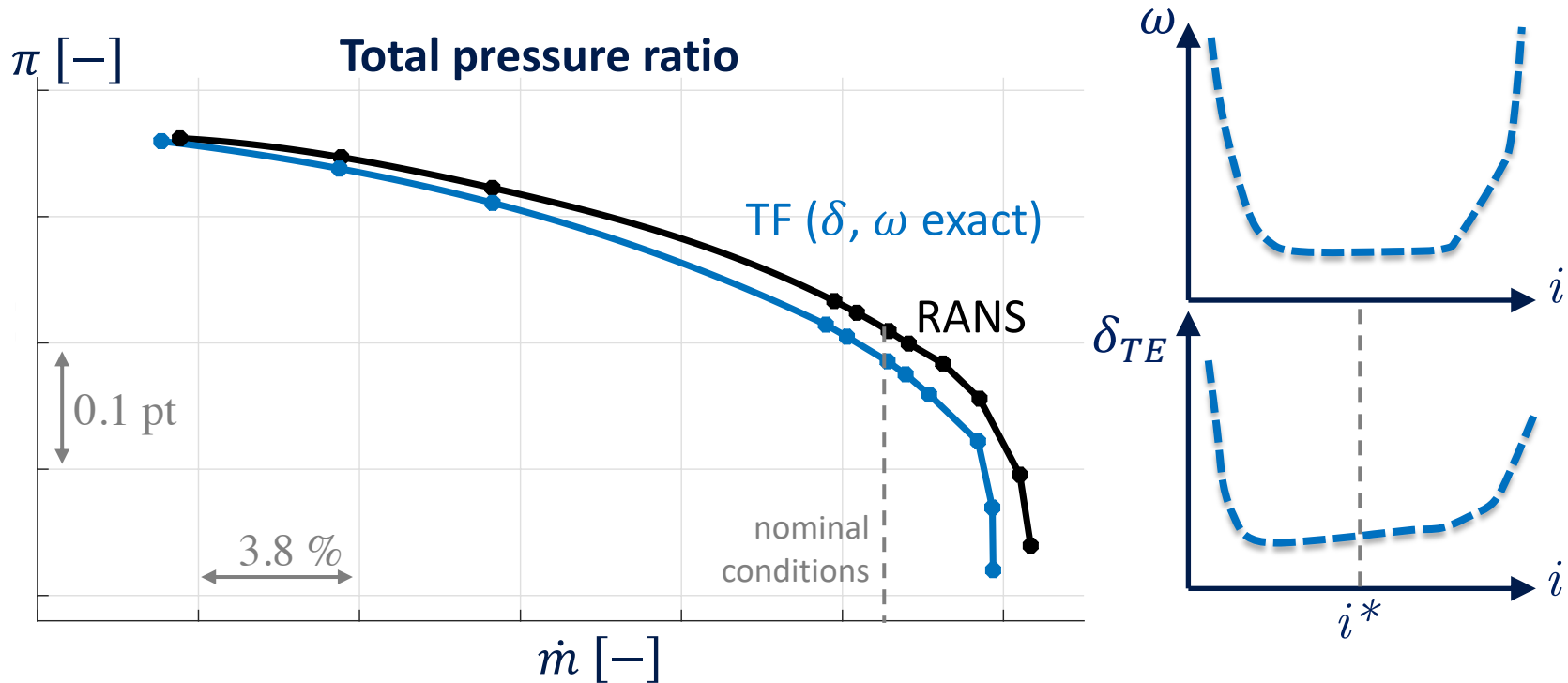


Model able to predict performance

But exact δ, ω unknown in practice...

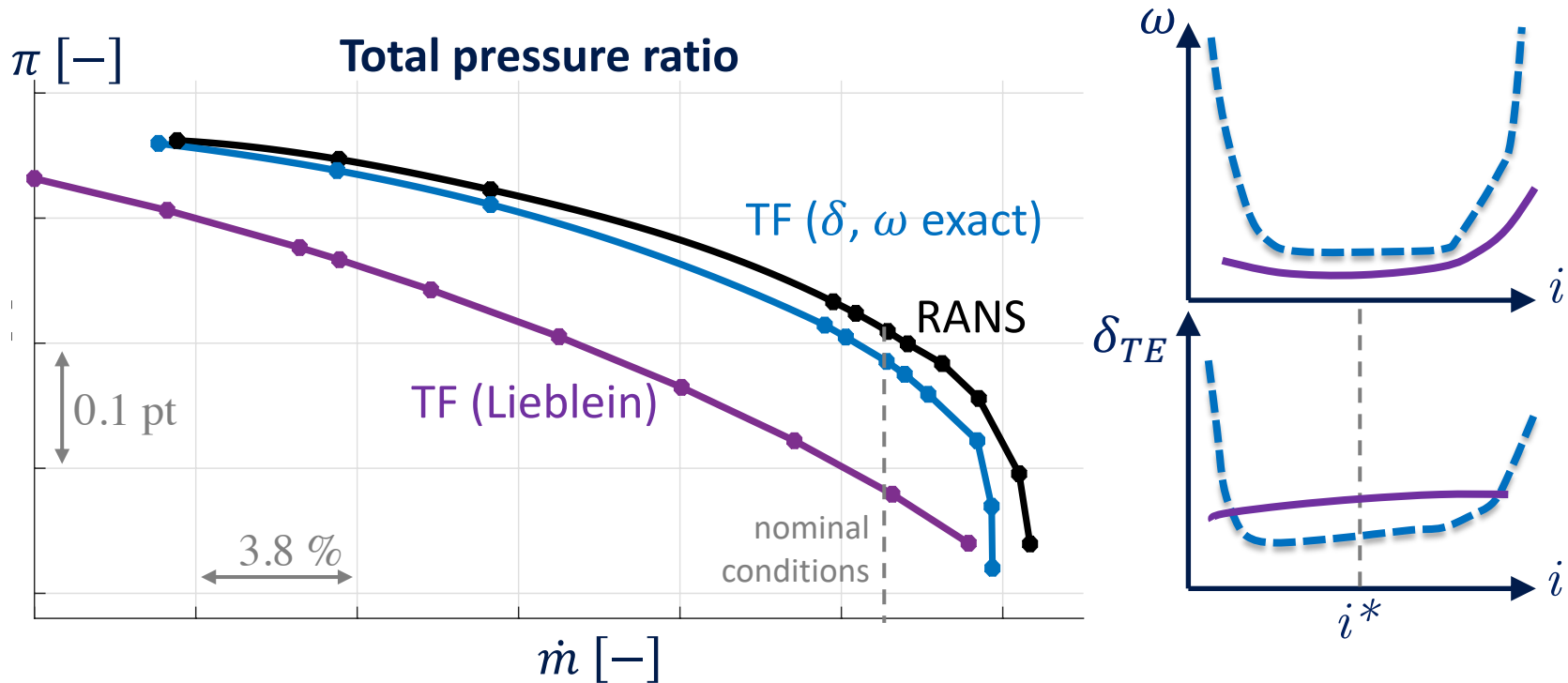
Correlations assessment

- Error quantification of correlations for δ , ω



Correlations assessment

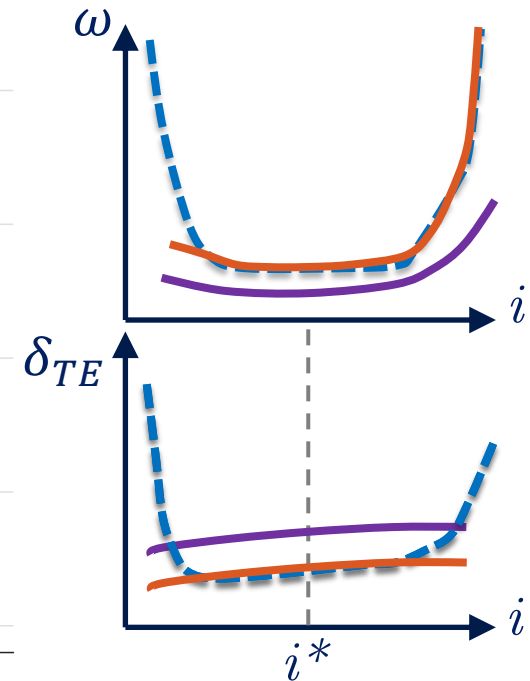
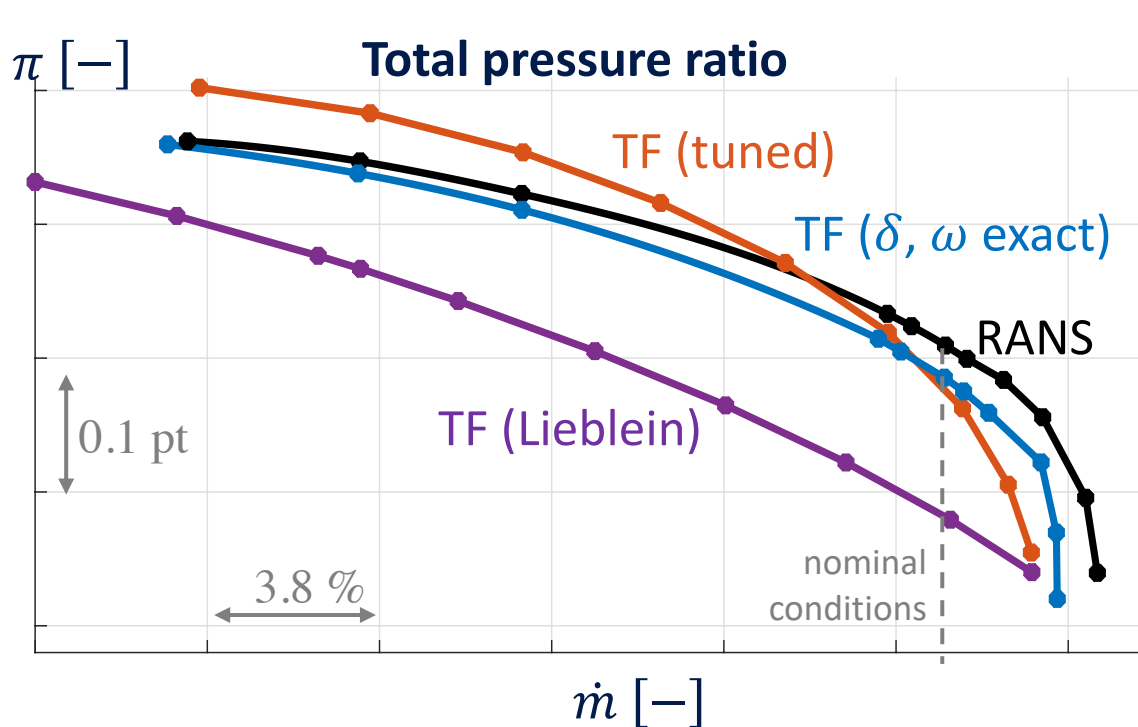
- Error quantification of correlations for δ , ω



- 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

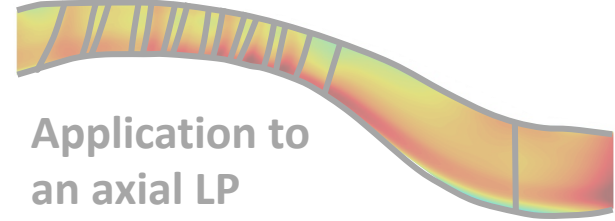
1

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

Geometrical variability



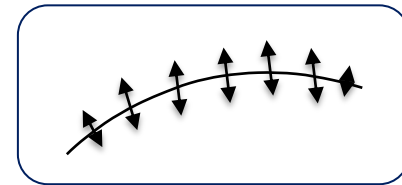
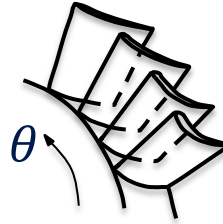
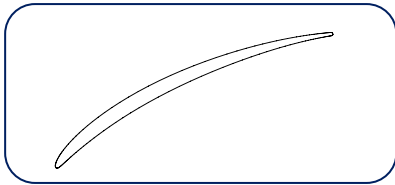
[SAB]

Geometry in through-flow model

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

θ -averaging

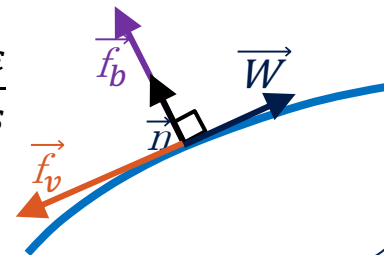
$$D_t \bar{U}(r, \theta, x) = \bar{G}(U, r, \theta, x)$$



**Camber line coordinates
&
Thickness distribution**

Blade forces

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



Direct
impact

Correlations

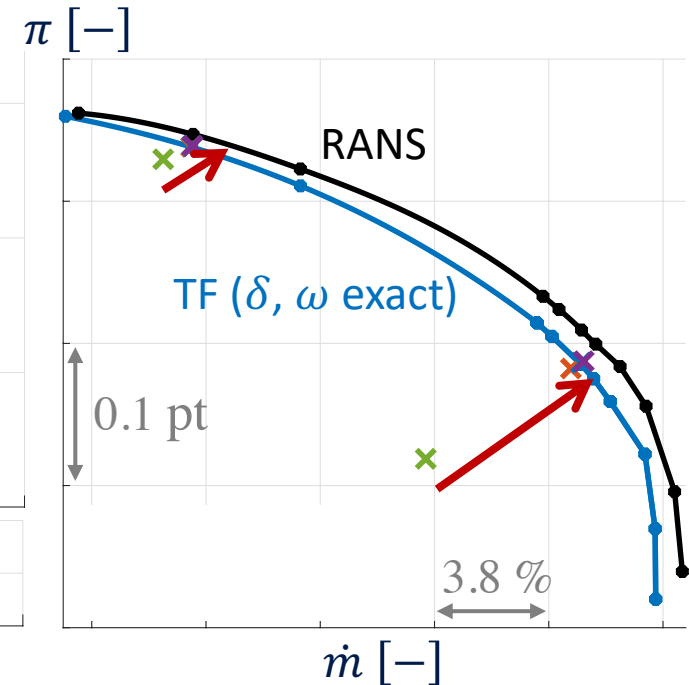
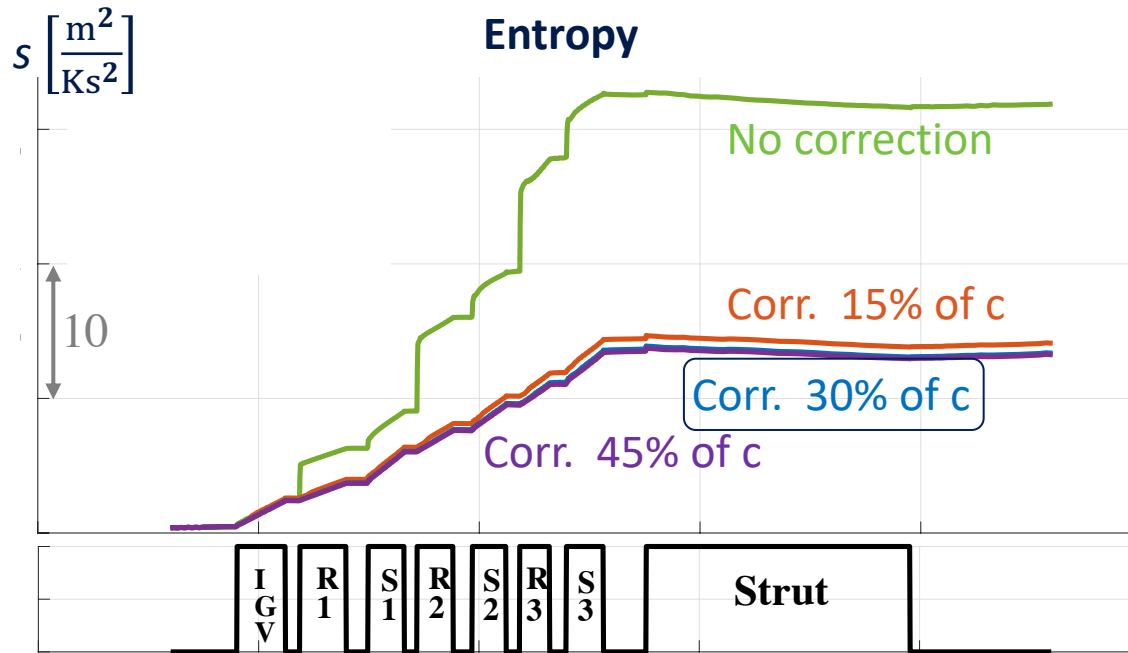
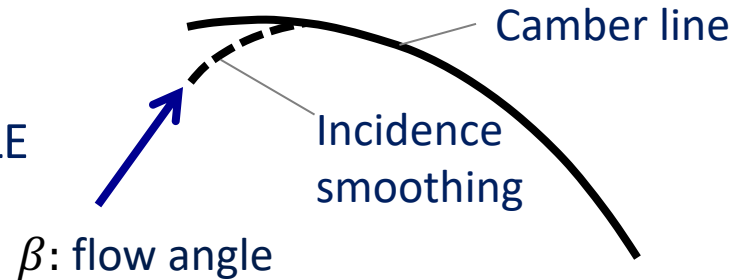
$$\delta_{TE}, \omega, i^* = f \left(\begin{array}{c} \text{geometry,} \\ \text{flow quantities} \end{array} \right)$$



Indirect
impact

Incidence correction

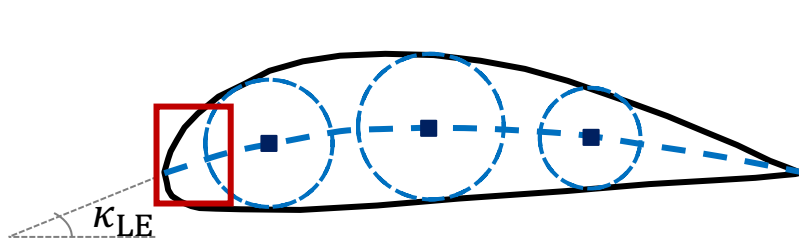
- Avoid flow angle discontinuity
- Modification of blade skeleton @ LE
- Unchanged correlation input



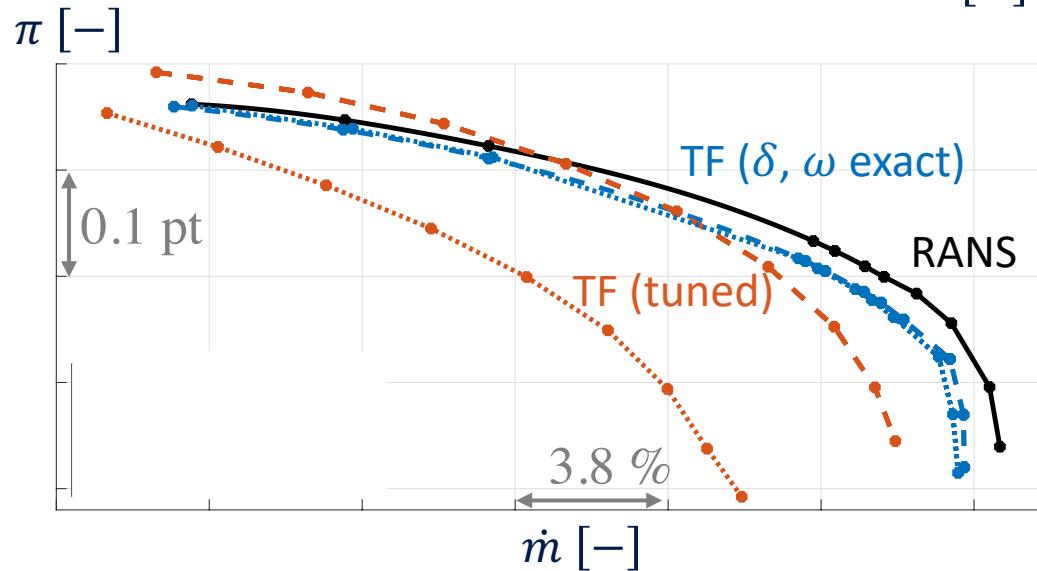
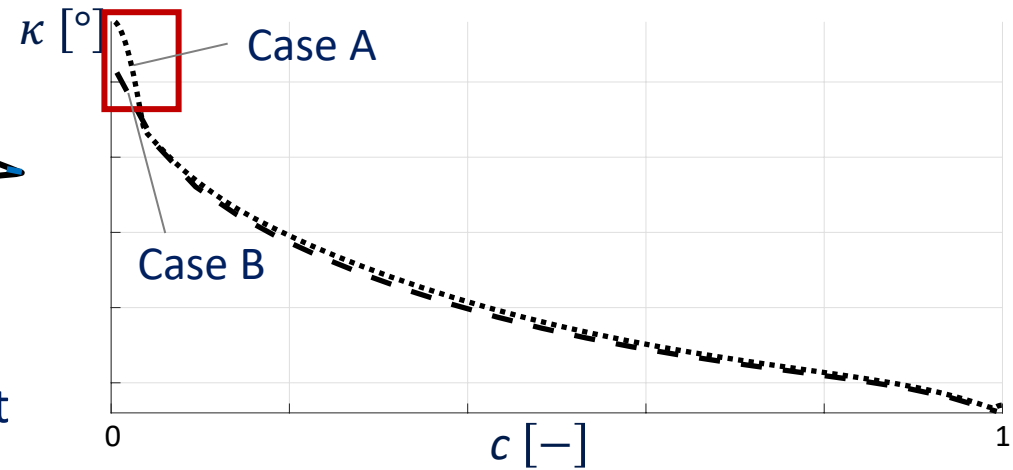
Incidence correction can smooth variability @ LE



Camber line definition

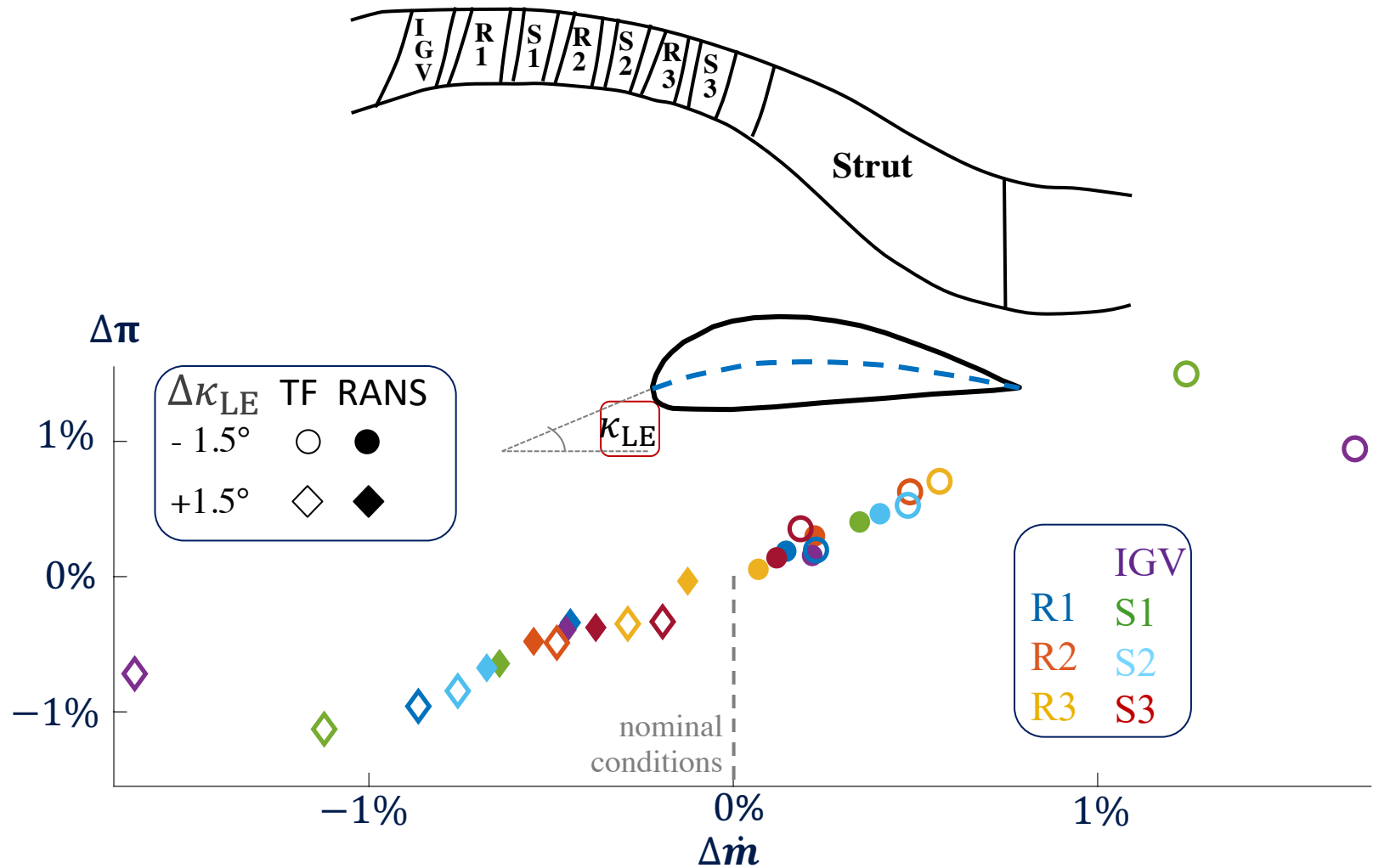


Definition not unique close to LE
→ Large impact on correlation input



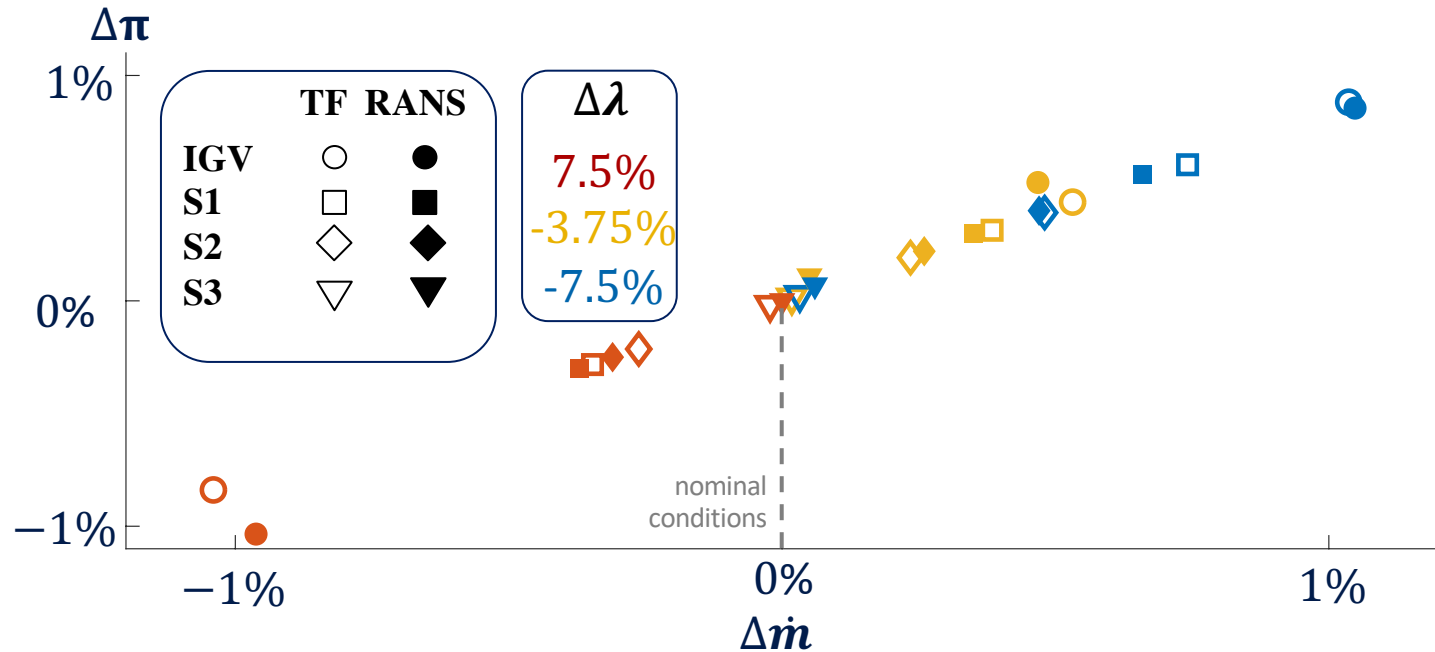
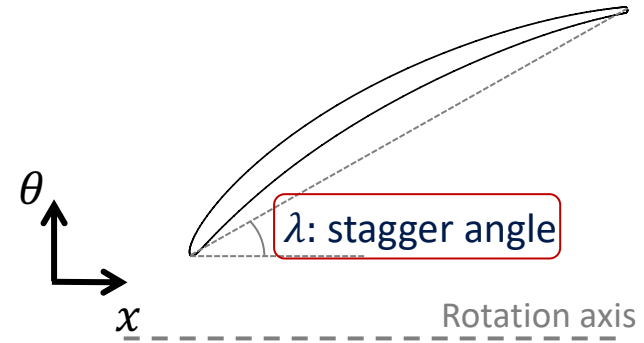
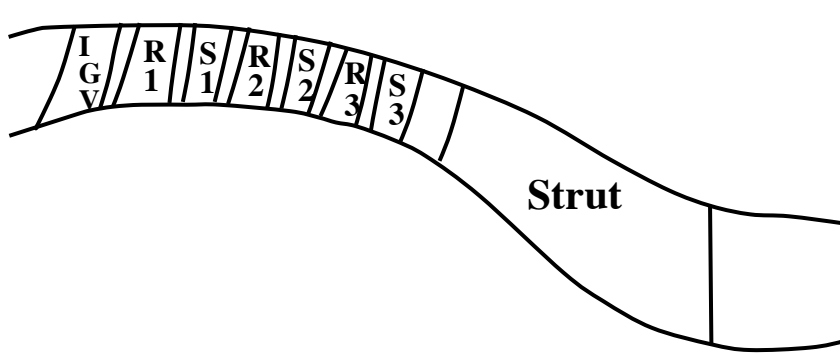
Strong dependence of model prediction with respect to LE blade angle

Applications: blade angle variability



Overestimation of performance variability

Applications: stagger angle variability



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

- Modeling aspects (incidence correction and camber line definition) **smear variability propagation @ LE**
- Global good agreement for other **performance variation**

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

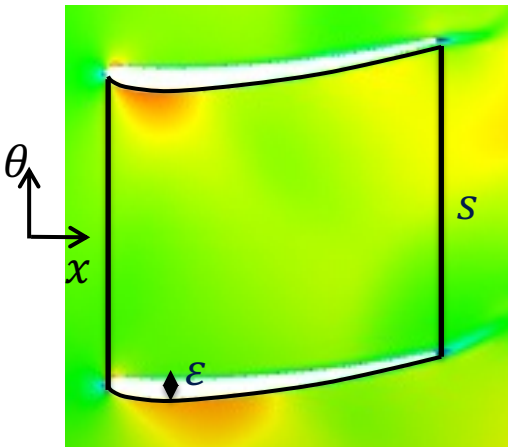


BACK-UP

Viscous through-flow model

Circumferential averaged Navier-Stokes equations:

Conservative variables



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

Blockage factor

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

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

Non-intrusive formulation for CFD solver:

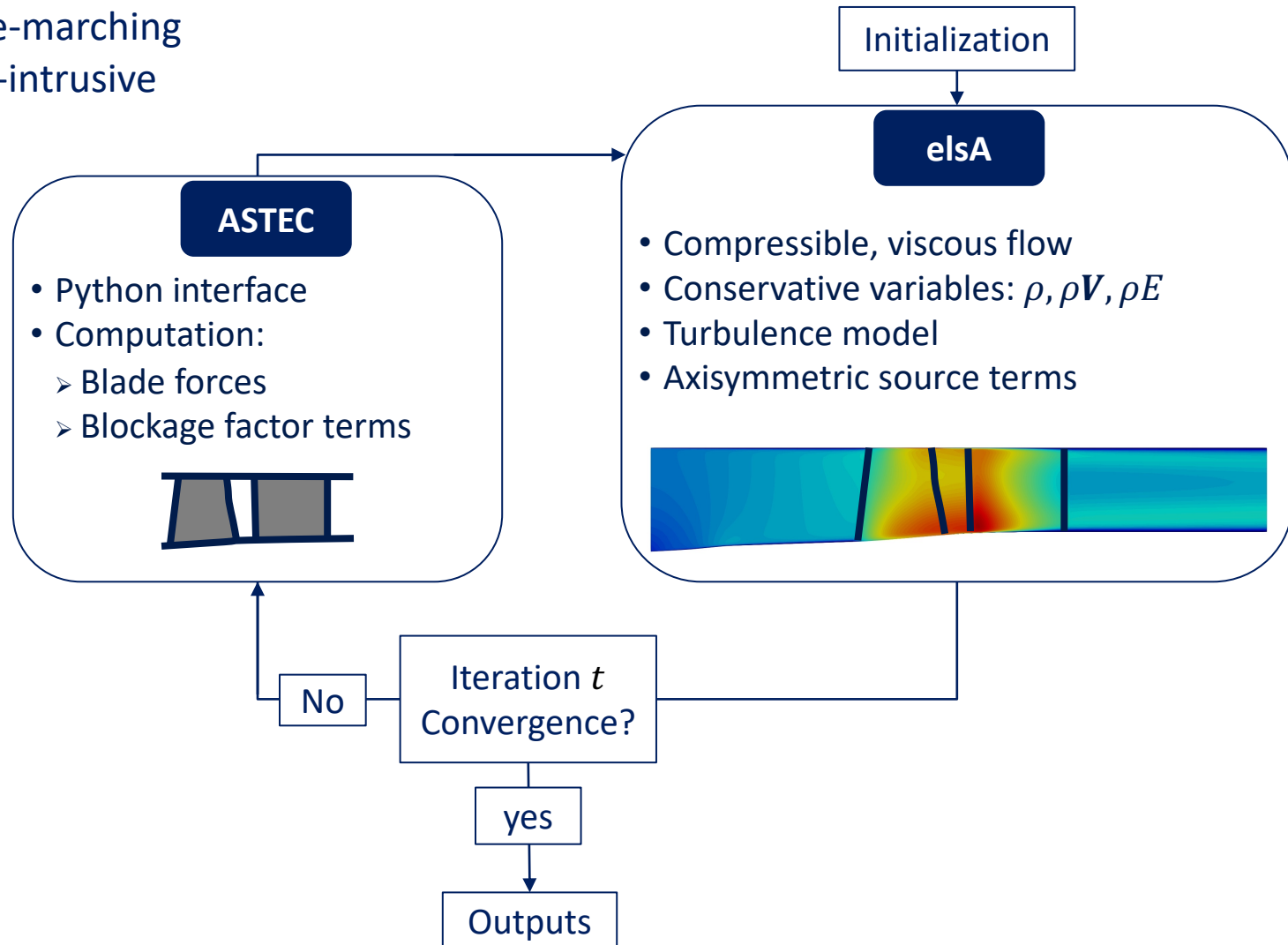
$$\frac{\partial U}{\partial t} + \frac{\partial (F - F_v)}{\partial x} + \frac{\partial (G - G_v)}{\partial r} = S + \boxed{\frac{(F_v - F)}{b} \frac{\partial b}{\partial x} + \frac{(G_v - G)}{b} \frac{\partial b}{\partial r}}$$

Blockage factor terms (known)

Viscous through-flow model: ASTEC

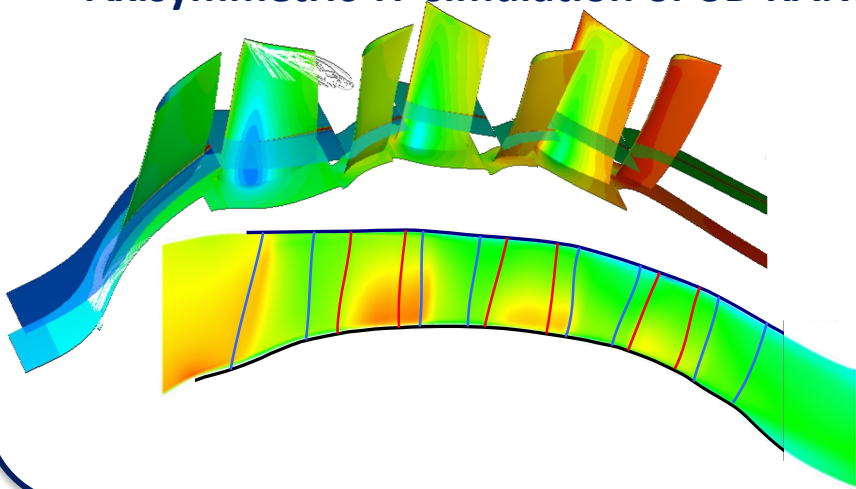
Methodology:

- Time-marching
- Non-intrusive



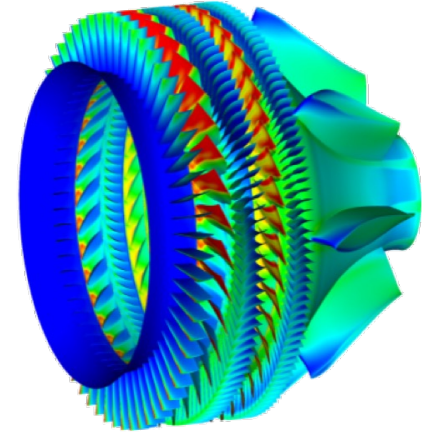
Relevance

Axisymmetric TF simulation or 3D RANS

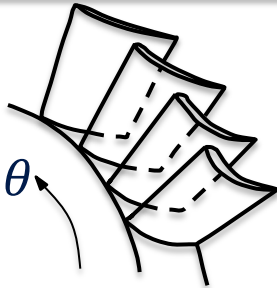


≠

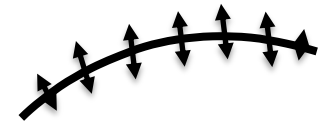
Mistuning



But



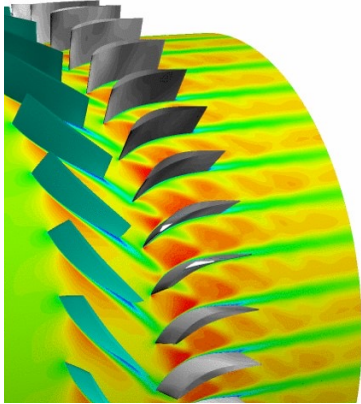
Averaged blade row



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

Correlations assessment

- Error quantification of **correlations** for δ , ω

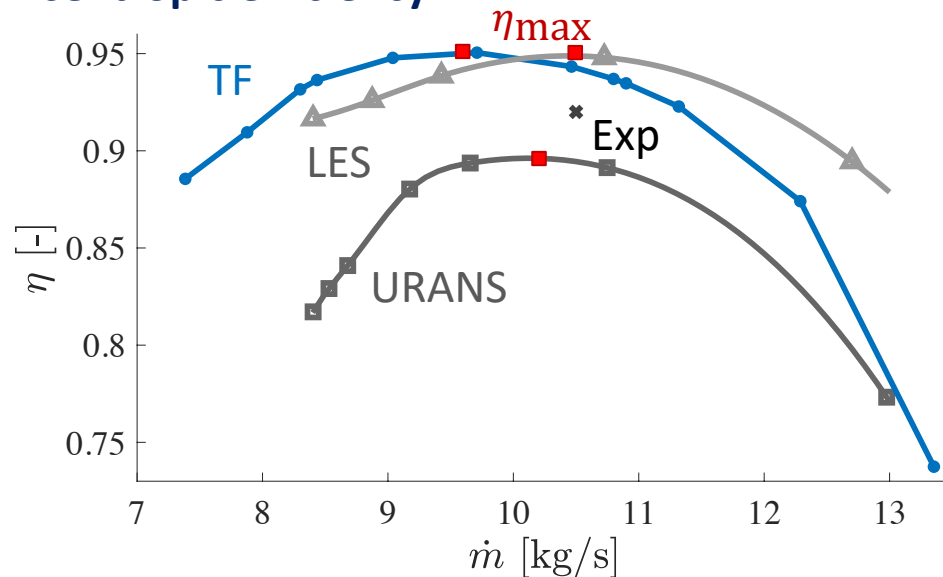


[Moreau 2019]

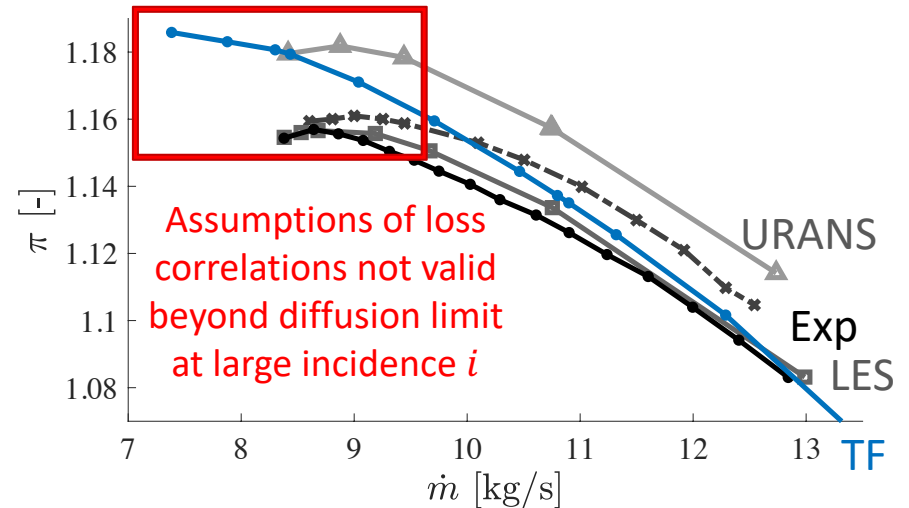
Test-case 2: CME2

- Research compressor designed by Safran Aircraft Engines
- Low speed flow
- Correlations calibrated at these conditions

Isentropic efficiency



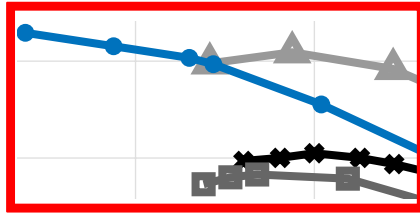
Total pressure ratio



Source: [Gourdain 2015]

CME2: Diffusion limit

Total pressure ratio

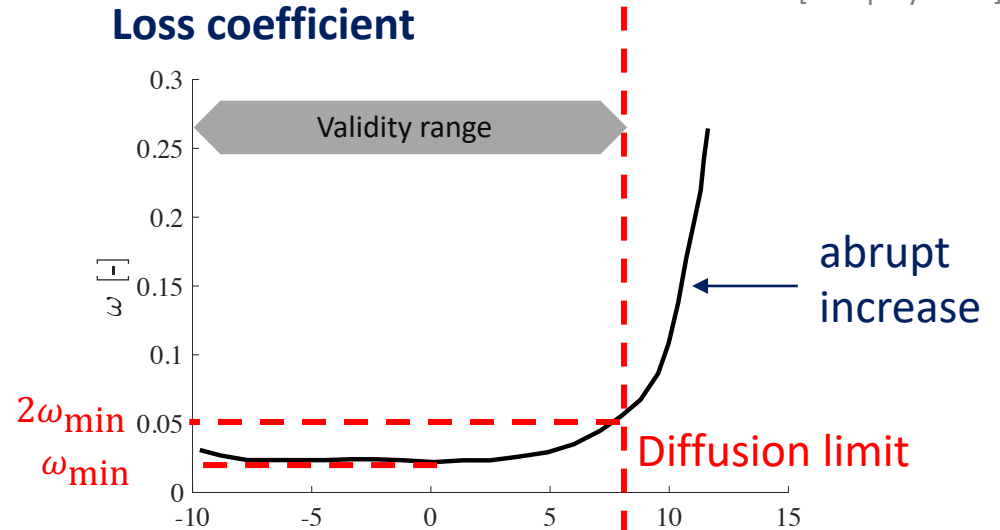


**Assumptions of loss correlations
not valid beyond diffusion limit
at large incidence i**

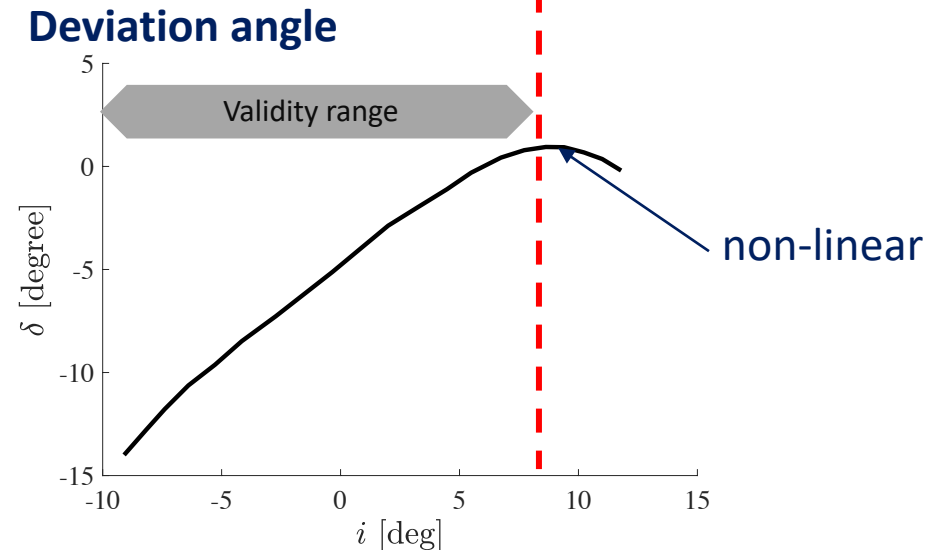
Measurements of C4-series cascade ($M = 0.4$)

[Cumpsty 1989]

Loss coefficient



Deviation angle



Modern compressor: comparison to RANS

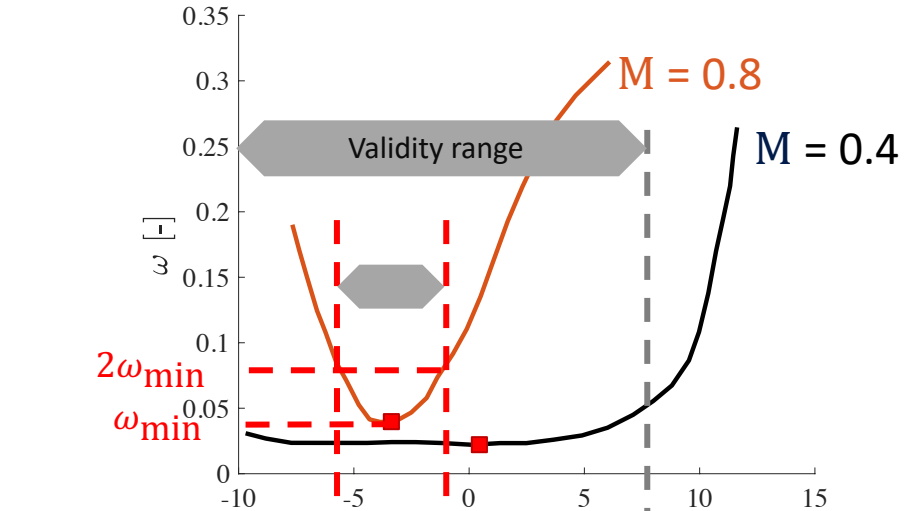
Impact of Mach number

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

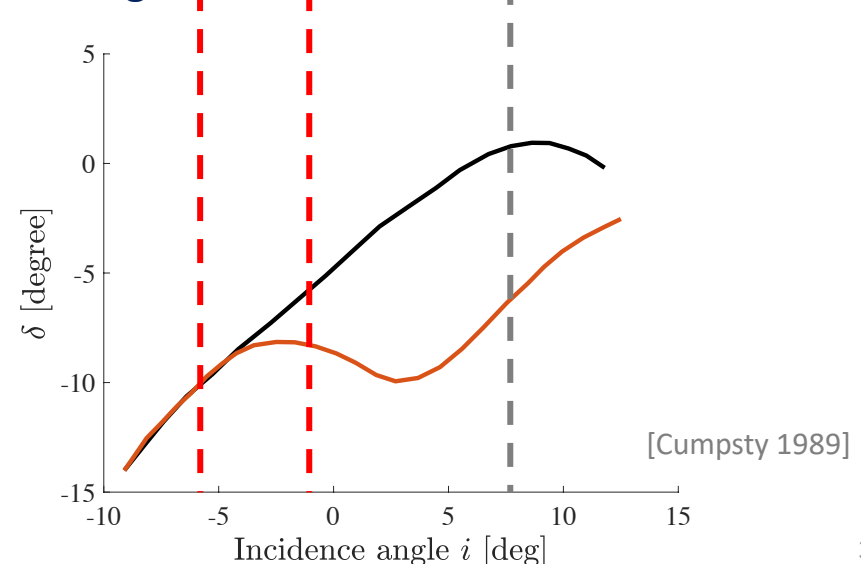
Correlations **not calibrated** for these flow conditions

Measurements of C4-series cascade

Loss coefficient



Deviation angle



[Cumpsty 1989]