

Explanation of Surface Deviations by Manufacturing Modes

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Towards a Fully-Automated Robust Design Procedure

Deviation Representation Compatible with Design Procedure

Decomposition into Manufacturing Modes

Application to a High-Pressure Turbine Blade

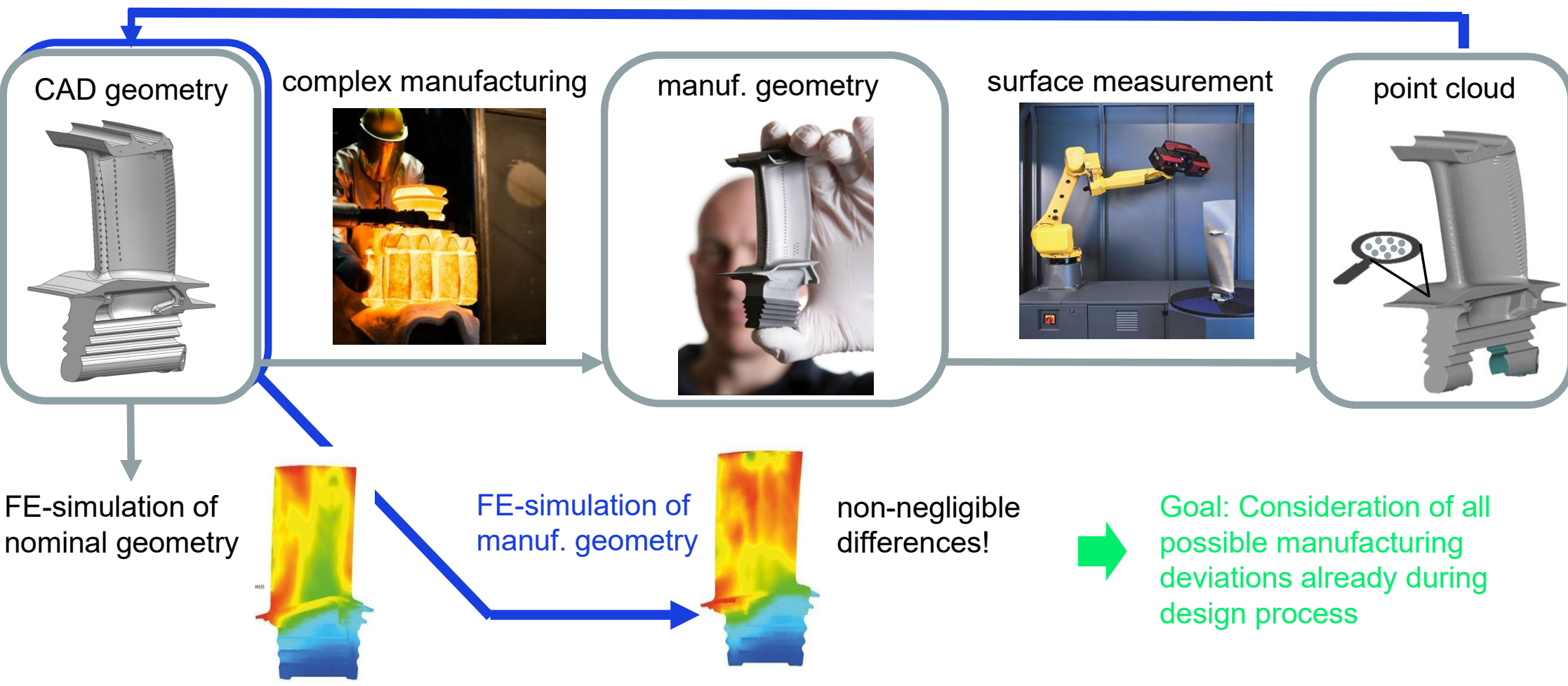


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Towards a Fully-Automated Robust Design Procedure

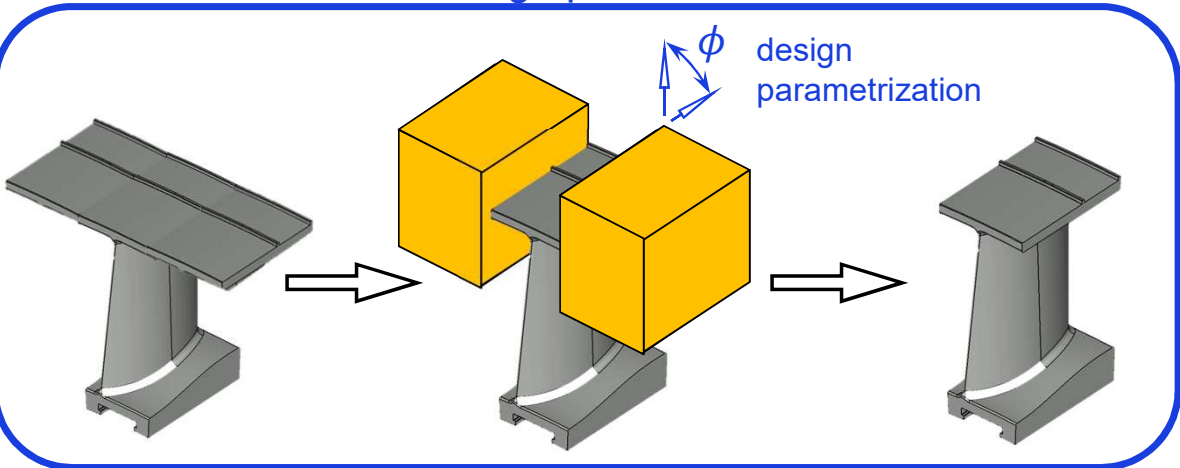
➤ Need for consideration of manufacturing variabilities



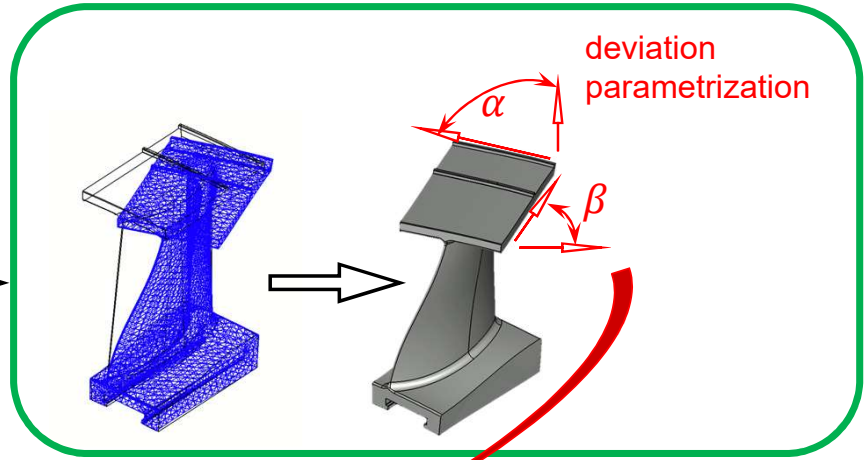
Goal: Consideration of all possible manufacturing deviations already during design process

- Current challenges for fully-automated robust design
 1. design parametrization \neq deviation parametrization

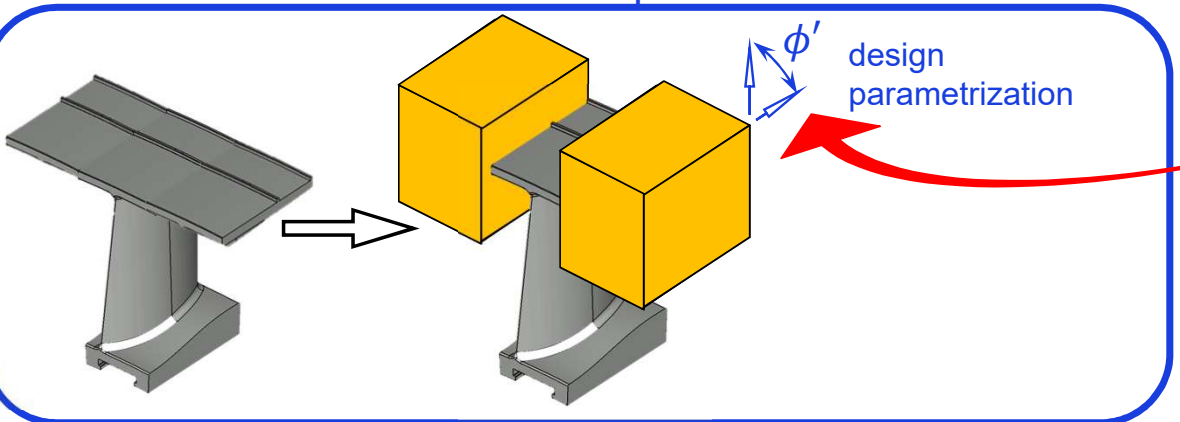
design procedure



deviation representation



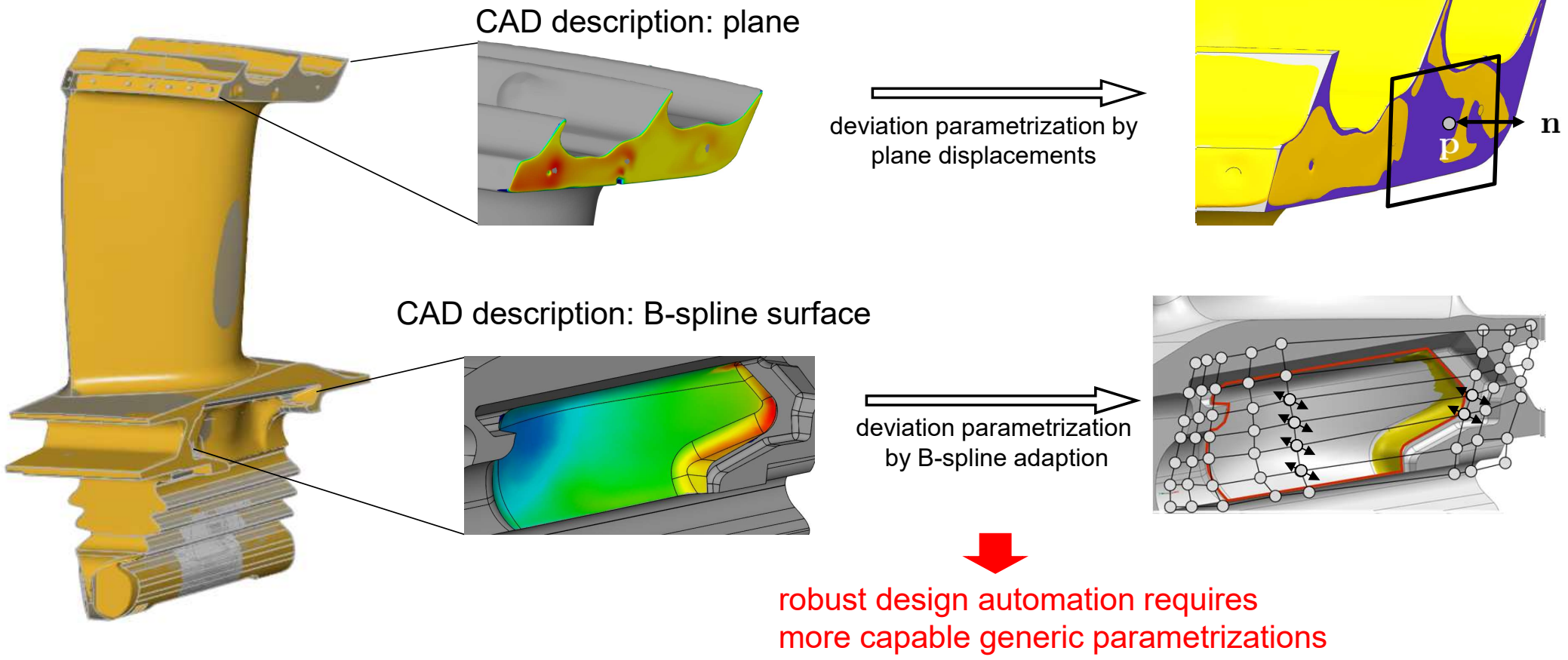
deviation replication



not compatible!

Integration of deviation representation into design procedure necessary

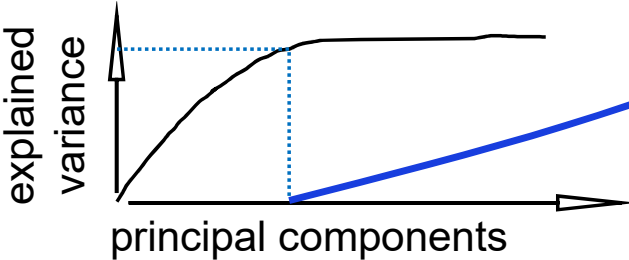
- Current challenges for fully-automated robust design
 1. design parametrization \neq deviation parametrization
 2. different features \rightarrow different deviation parametrizations



- Current challenges for fully-automated robust design
 1. design parametrization \neq deviation parametrization
 2. different features \rightarrow different deviation parametrizations
 3. multiple measurements \rightarrow multiple stochastic parameters

common practice:

- i. collect deviation vectors and perform PCA
- ii. consider principal components for highest explained variance



typical reductions:
637 \rightarrow 100 components (Beck, et. al, 2019)
80 \rightarrow 15 components (Voigt, et. al, 2018)

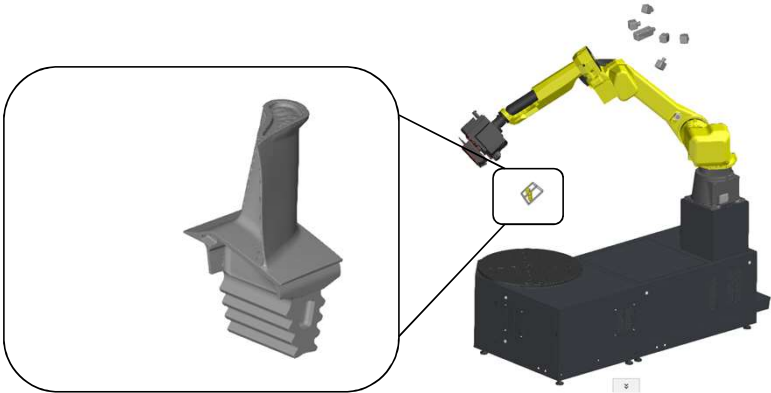


efficient robust optimization requires far fewer uncertain parameters

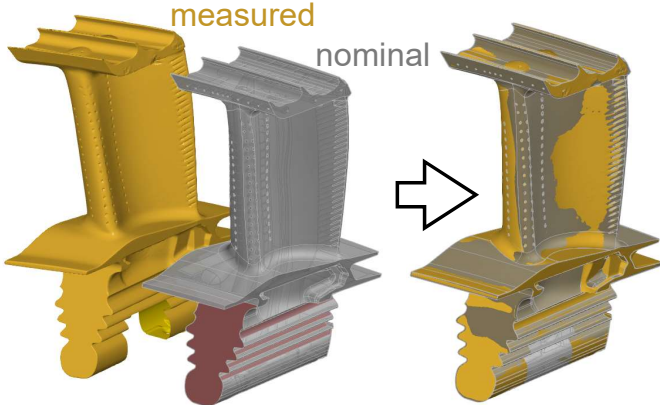
Deviation Representation Compatible with Design Procedure

➤ Preparation of surface measurements

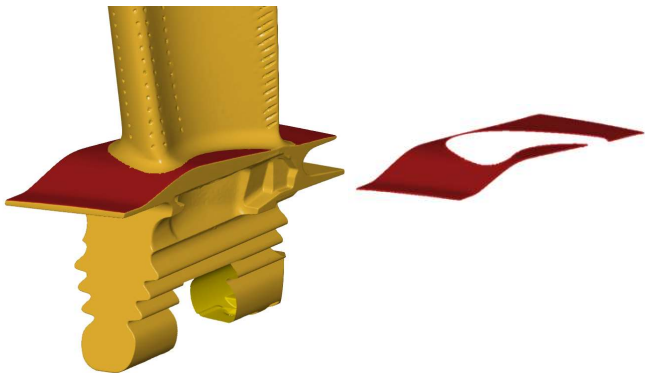
1. Surface measurement



2. Alignment to nominal geometry

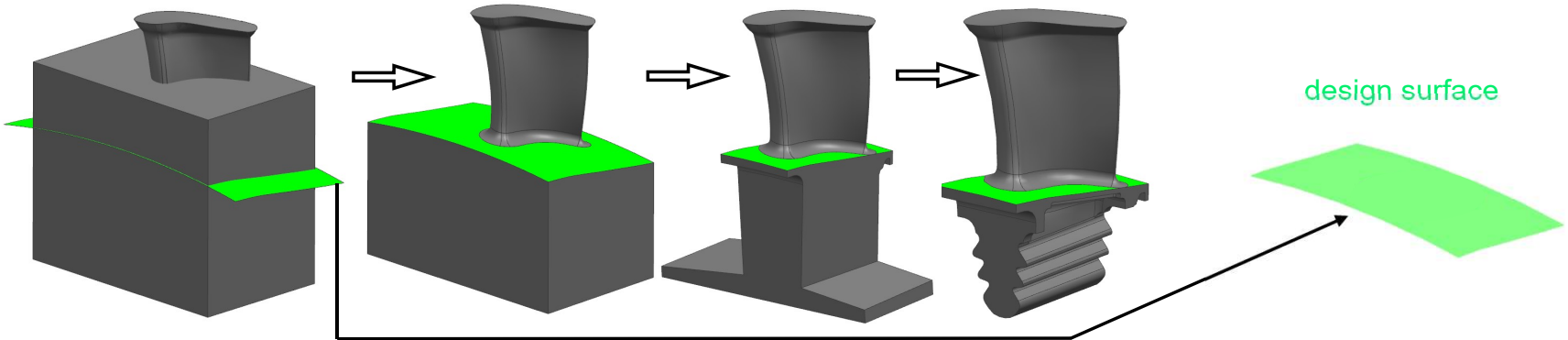


3. Segmentation of surface of interest

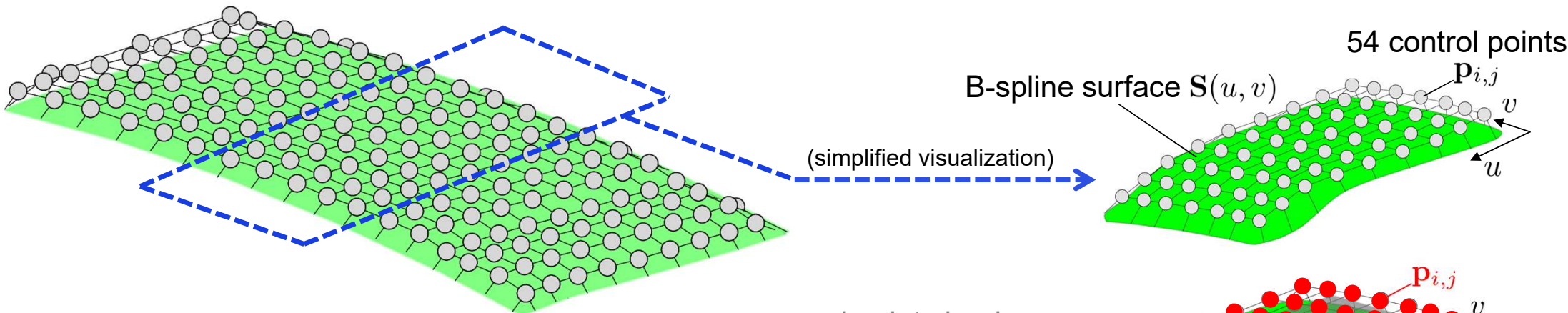


➤ Preparation of CAD surfaces

1. Extraction of design surface from design procedure



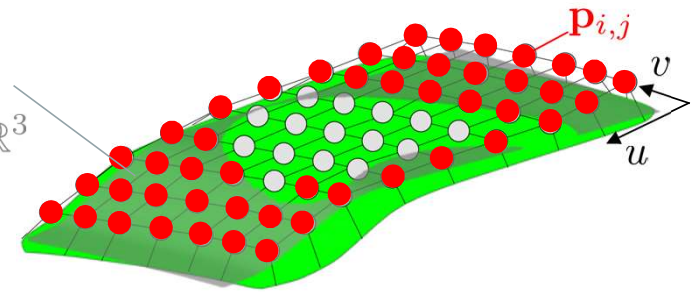
2. Re-parametrization of **design surface** as B-spline surface with increased number of control points



3. Selection of relevant control points

$$\mathcal{I} = \left\{ (i, j) \in [0, n] \times [0, m] \subset \mathbb{N}^2 \mid \exists \mathbf{r} \in \mathcal{M} : \|\mathbf{p}_{i,j} - \mathbf{r}\| \leq \delta_{\mathcal{I}} \right\}$$

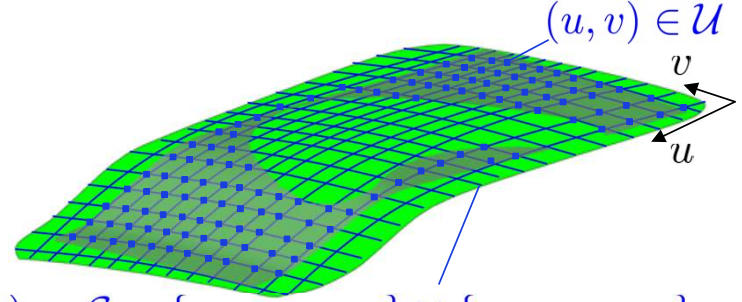
measured point cloud
 $\mathcal{M} = \{\mathbf{r}_1, \dots, \mathbf{r}_M\}, \mathbf{r} \in \mathbb{R}^3$



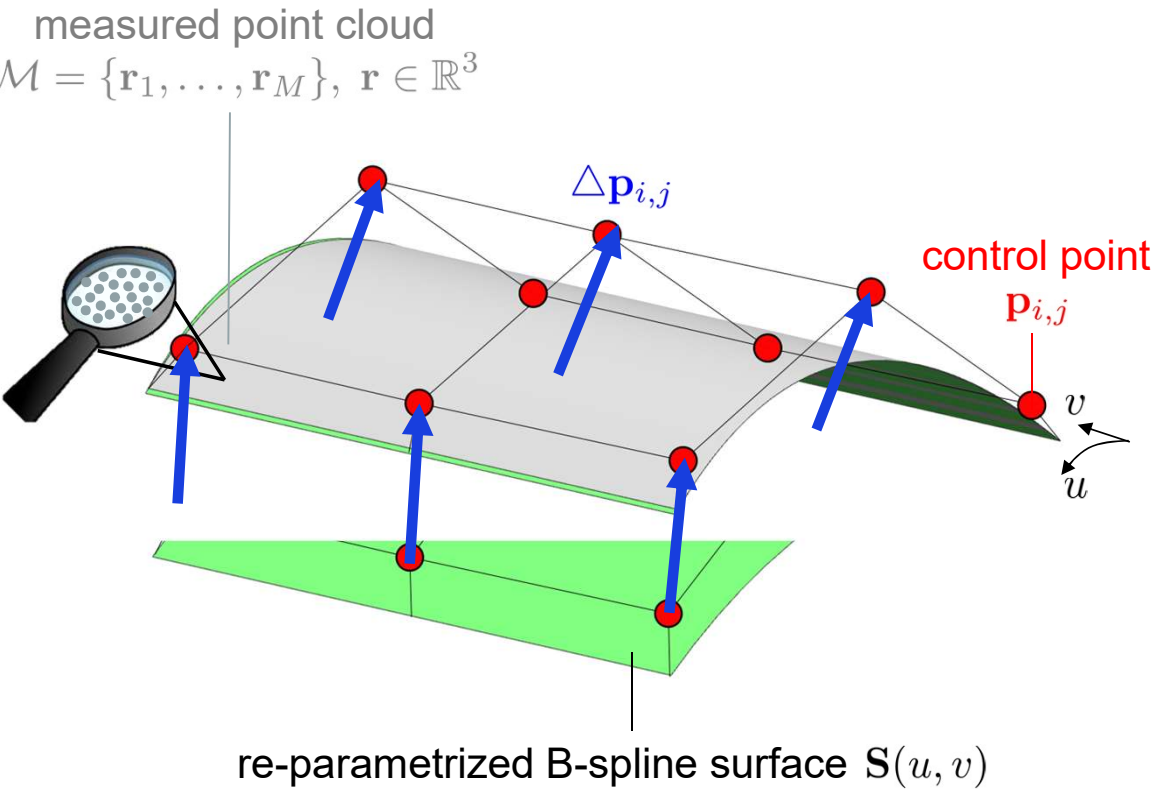
4. Association of B-spline parameters (u, v) with point cloud

$$k(u, v) = \operatorname{argmin}_{\kappa: \mathbf{r}_{\kappa} \in \mathcal{M}} \|\mathbf{S}(u, v) - \mathbf{r}_{\kappa}\|, \quad \mathcal{U} = \left\{ (u, v) \in \mathcal{G} \mid \|\mathbf{S}(u, v) - \mathbf{r}_{k(u,v)}\| \leq \delta_{\mathcal{U}} \right\}$$

fine grid on $\mathbf{S}(u, v)$: $\mathcal{G} = \{u_1, \dots, u_{n'}\} \times \{v_1, \dots, v_{m'}\}$



➤ B-spline morphing using surface measurements



re-parametrized B-spline surface

$$\mathbf{S}(u, v) = \sum_{i=0}^n \sum_{j=0}^m N_i^q(u) N_j^q(v) \mathbf{p}_{i,j}$$

B-spline morphing formulation

$$\hat{\mathbf{S}}(u, v) = \sum_{i=0}^n \sum_{j=0}^m N_i^q(u) N_j^q(v) (\mathbf{p}_{i,j} + \Delta \mathbf{p}_{i,j}) \stackrel{!}{=} \mathbf{r}_{k(u,v)}$$

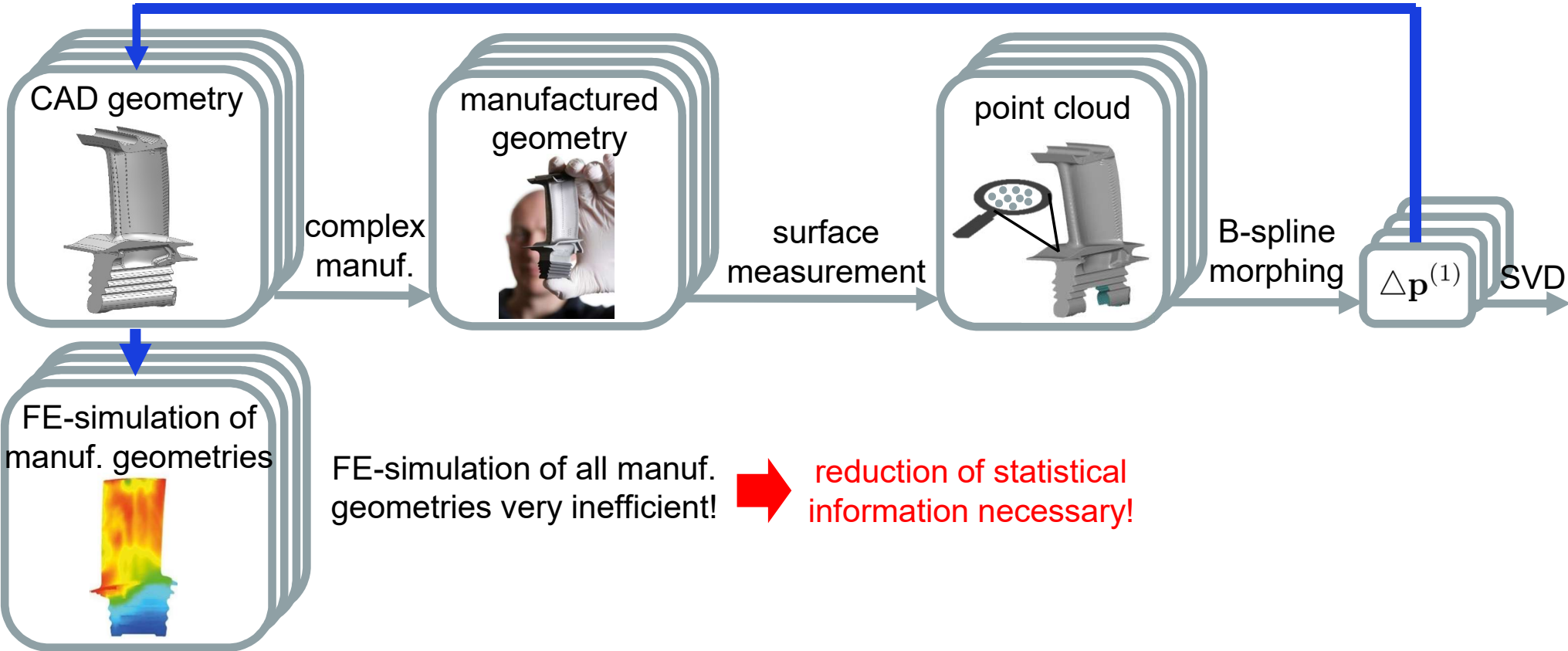
$$\sum_{(i,j) \in \mathcal{I}} N_i^q(u) N_j^q(v) \Delta \mathbf{p}_{i,j} = \mathbf{r}_{k(u,v)} - \mathbf{S}(u, v)$$

$$\mathbf{NP} = \mathbf{D}$$

solution of the overdetermined system of linear equations with linear regression

Decomposition into Manufacturing Modes

➤ Analysis of multiple deviating surfaces is computationally inefficient



➤ Computation of manufacturing modes (Urbano et.al., 2019):

- Collect control point deviations of scan population to construct data matrix:

$$\Delta \mathbf{p}^{(i)} = [\dots \Delta p_{i,j}^x \dots | \dots \Delta p_{i,j}^y \dots | \dots \Delta p_{i,j}^z \dots]^\top, \quad \Delta \bar{\mathbf{p}} = \frac{1}{S} \sum_s^S \Delta \mathbf{p}^{(s)},$$

$$\mathbf{X} = [\Delta \mathbf{p}^{(1)} - \Delta \bar{\mathbf{p}}, \dots, \Delta \mathbf{p}^{(S)} - \Delta \bar{\mathbf{p}}].$$

- Perform singular-value decomposition of data matrix:

$$\mathbf{X} = \sum_{i=1}^r \sigma_i \mathbf{u}_i \mathbf{v}_i^\top,$$

for arbitrary surface deviations

$$\mathbf{x} = \mathbf{X}\boldsymbol{\alpha} = (\mathbf{U}\boldsymbol{\Sigma}\mathbf{V}^\top) \boldsymbol{\alpha} = \sum_{i=1}^r (\sigma_i \mathbf{u}_i \mathbf{v}_i^\top) \boldsymbol{\alpha}$$

$$\equiv \sum_{i=1}^r (\sigma_i \mathbf{v}_i^\top \boldsymbol{\alpha}) \mathbf{u}_i =: \sum_{i=1}^r \beta_i \mathbf{u}_i \quad \text{uncorrelated manuf. modes}$$

- Projection of control point displacements $\Delta \mathbf{q}$ of arbitrary surface scan

$$\beta_i = \mathbf{u}_i^\top (\Delta \mathbf{q} - \Delta \bar{\mathbf{p}}) \quad (\text{mode amplitude})$$

$$\Delta \mathbf{q} = \Delta \bar{\mathbf{p}} + \beta_1 \mathbf{u}_1 + \beta_2 \mathbf{u}_2 + \dots + \beta_r \mathbf{u}_r$$

➤ Computation of an accurate low-order truncation:

- Compute reconstruction error $e_R^{(s),t}$ for **surface scan** (s) and **truncation order** t :

i. Approximate control point displacements: $\mathbf{x}^{(s),t} := \Delta \bar{\mathbf{p}} + \sum_{i=1}^t \beta_i^{(s)} \mathbf{u}_i \rightarrow \Delta \mathbf{p}_{i,j}^{(s),t}$ for $t < r$

ii. Create partially reconstructed surface: $\hat{\mathbf{S}}^{(s),t}(u,v) := \mathbf{S}(u,v) + \sum_{(i,j) \in \mathcal{I}} N_i^q(u) N_j^q(v) \Delta \mathbf{p}_{i,j}^{(s),t}$

iii. Compare with morphed surface: $e_R^{(s),t}(u,v) := \left| \left(\hat{\mathbf{S}}^{(s)}(u,v) - \hat{\mathbf{S}}^{(s),t}(u,v) \right)^\top \mathbf{n}(u,v) \right|$

- Obtain observable manufacturing modes:

$$t^* = \min t \text{ s.t. } \bar{e}_R(t) \leq \underbrace{e_S + \bar{e}_C + \bar{e}_M}_{\text{process uncertainty}}$$

where:

- e_S : measurement error,
- \bar{e}_C : re-parametrization error,
- \bar{e}_M : morphing error.

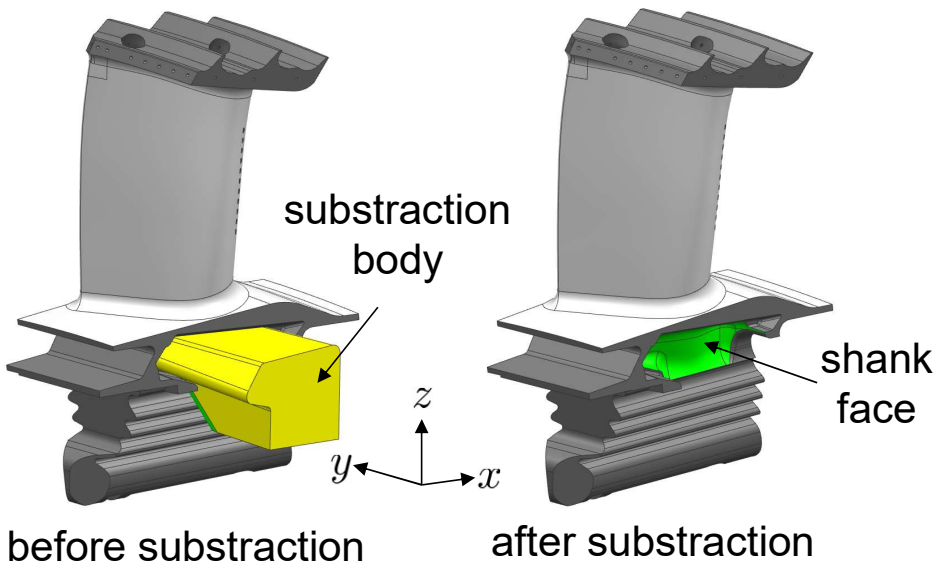
Re-parametrization e_C , morphing e_M and reconstruction e_R errors evaluated locally at each pair $(u,v) \in \mathcal{U}$.



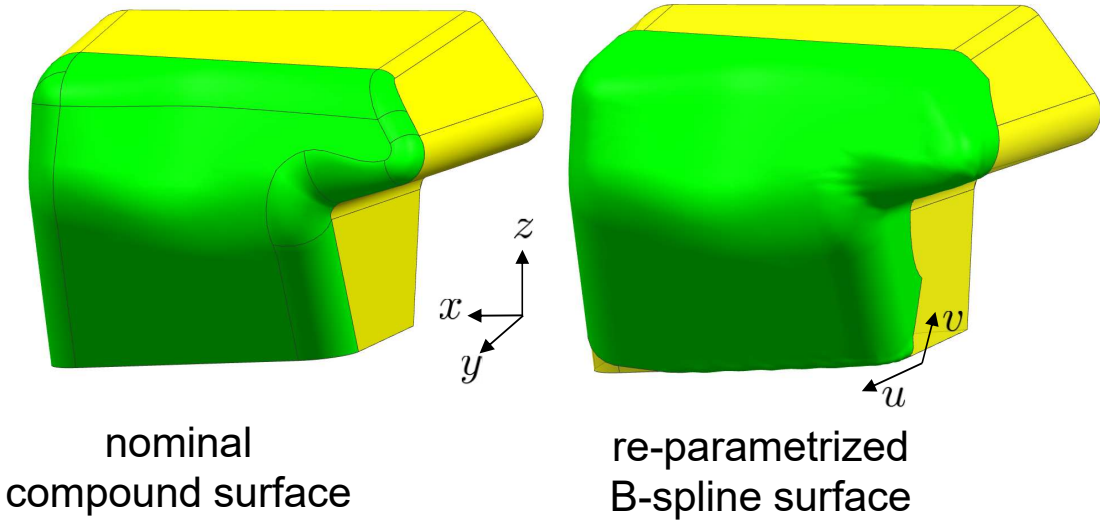
Obtain corresponding global errors $\bar{e}_C, \bar{e}_M, \bar{e}_R$ by taking percentiles over pairs $(u,v) \in \mathcal{U}$ and surface measurements

Application to the Shank Face of a High-Pressure Turbine Blade

➤ Design process of the shank face:

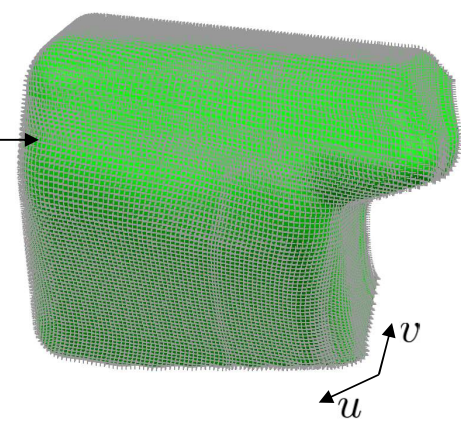


➤ Re-parametrization of the shank face:

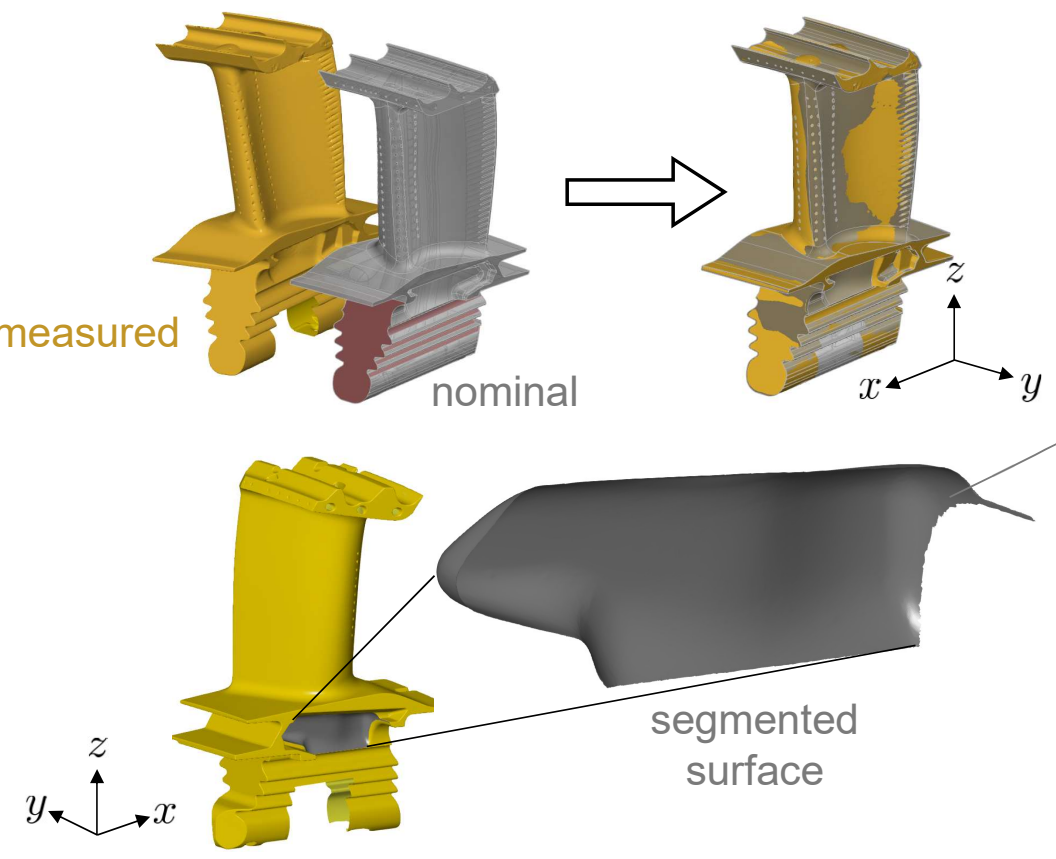


127 X 98
control points

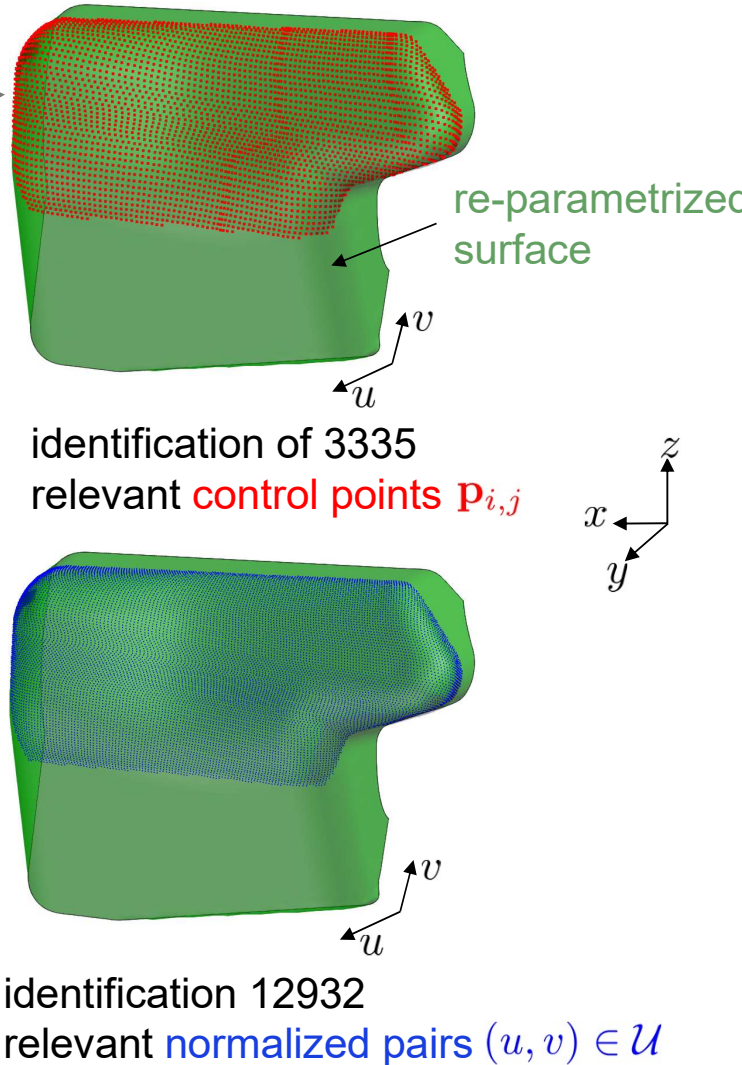
maximum re-parametrization
error $\bar{e}_C = 2 \mu\text{m}$



➤ Alignment and segmentation of the surface scans:



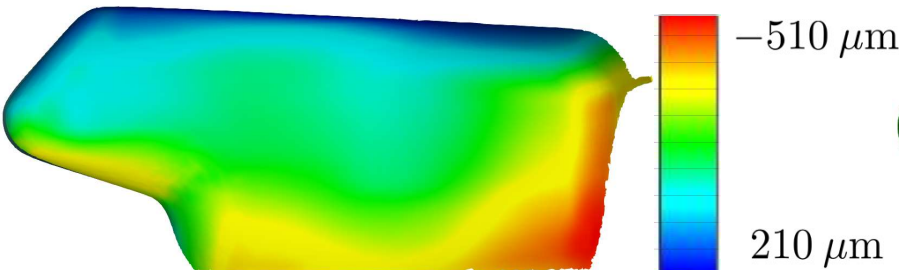
➤ Preparation of the re-parametrized surface:



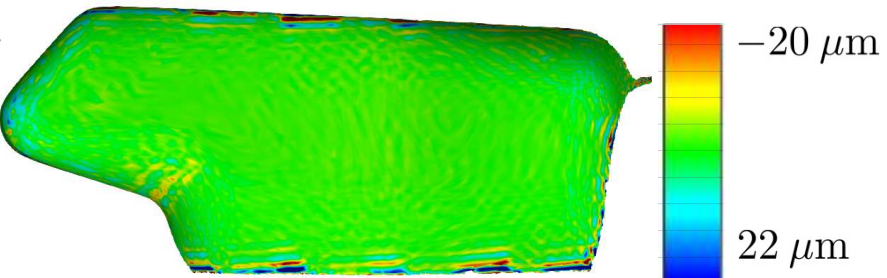
The authors thank the TU Dresden for the preparation and provision of the optical geometry measurements used in this investigation

➤ Morphing of 57 surface scans:

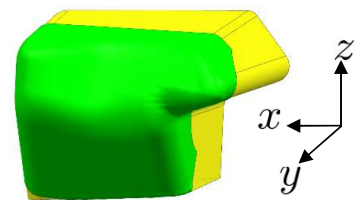
y-deviation, nominal - scan



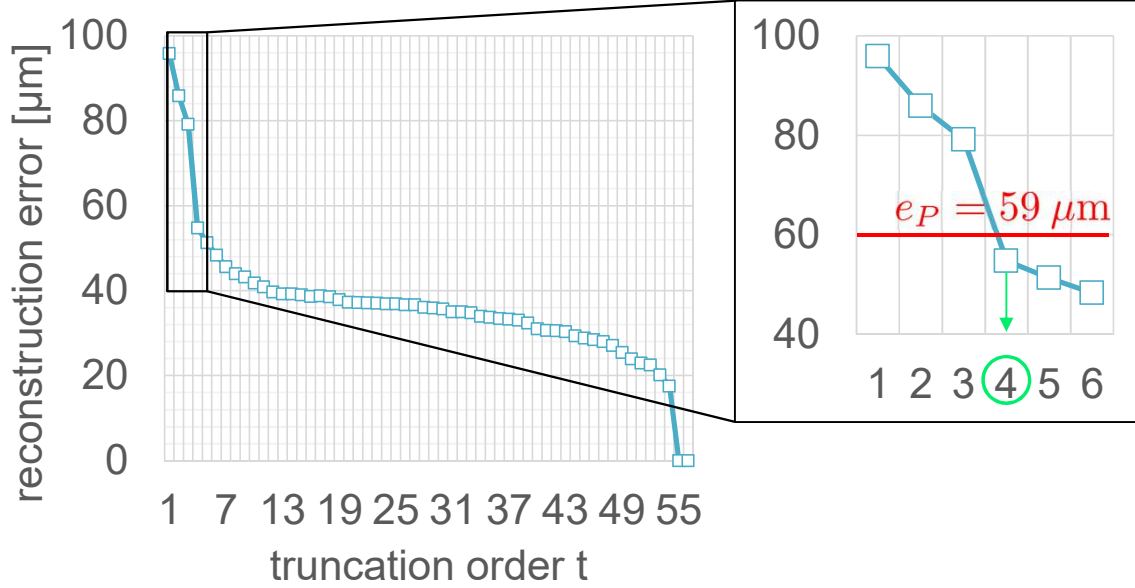
y-deviation, morphing - scan



sign convention based on subtraction body



➤ SVD and global reconstruction error:



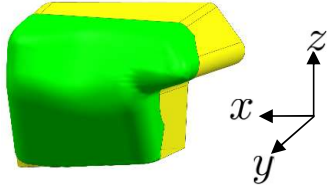
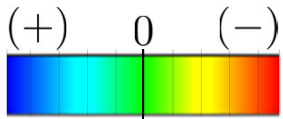
➤ Calculation of process uncertainty:

measurement error (joint precision and accuracy estimations)	$e_S = 35 \mu\text{m}$
re-parametrization error	$\bar{e}_C = 2 \mu\text{m}$
morphing error	$+ \bar{e}_M = 22 \mu\text{m}$
process error	$e_P = 59 \mu\text{m}$

↓
only 4 observable manufacturing modes!

➤ Analysis of manufacturing modes:

y-deviation from nominal

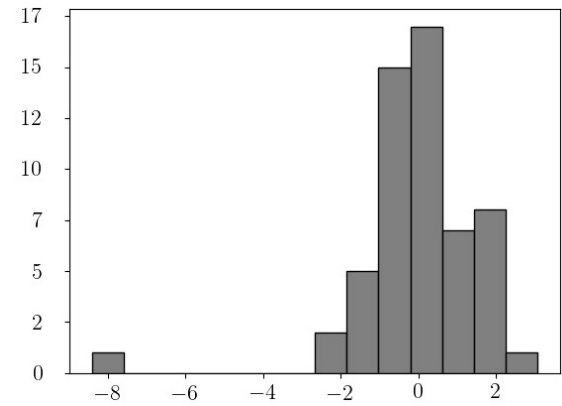
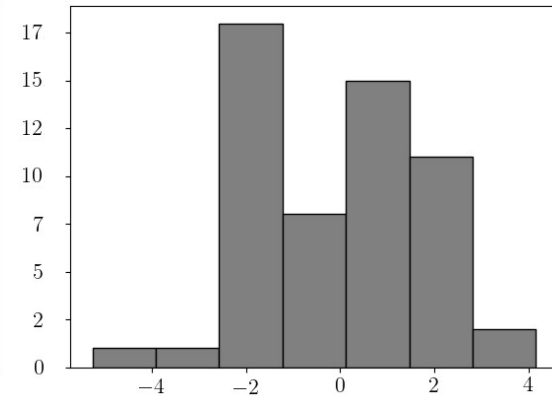
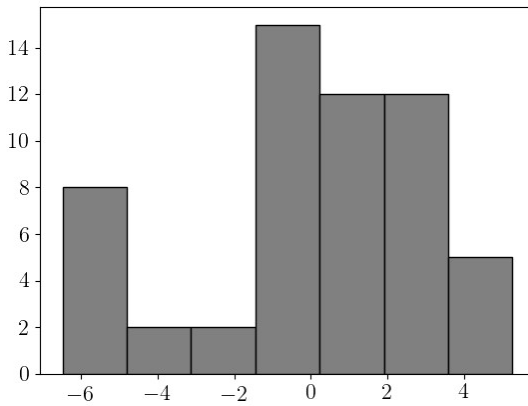
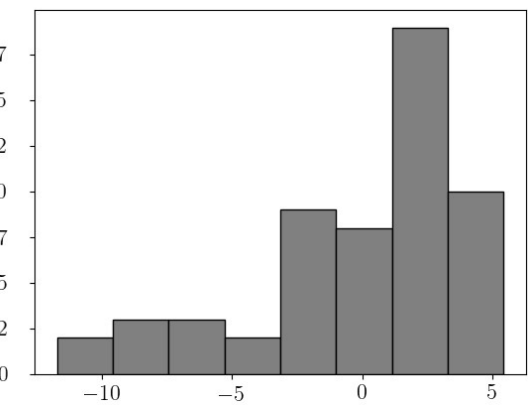
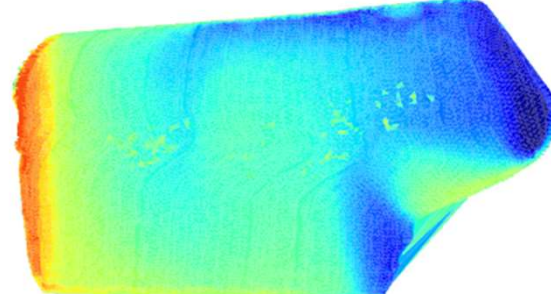
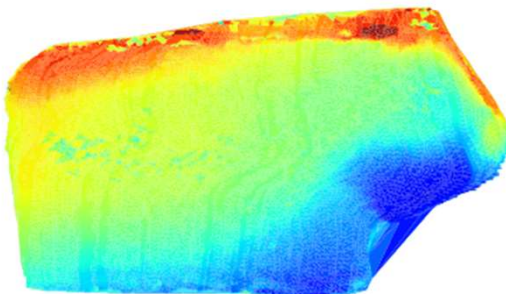
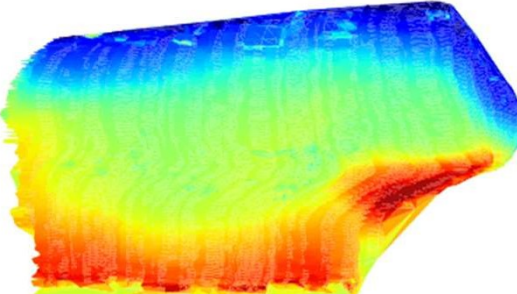
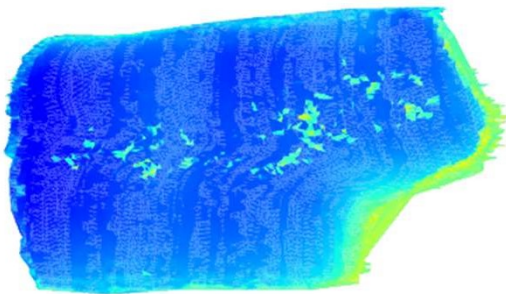


mode 1

mode 2

mode 3

mode 4



mode amplitude β_1

mode amplitude β_2

mode amplitude β_3

mode amplitude β_4

Conclusions

1. Re-parametrization of design surfaces as B-spline surfaces with high number of control points facilitates representation of complex manufacturing deviations and may be easily integrated into the design process.
2. Surface deviations of arbitrary features may be captured using control point displacements, which may be easily obtained from linear regression. On going work considers automatic B-spline re-parametrization of any geometric feature.
3. Singular value decomposition of control point displacements produces uncorrelated manufacturing modes. Furthermore, by using the information threshold a dramatic reduction in statistical information may be obtained.

References

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