

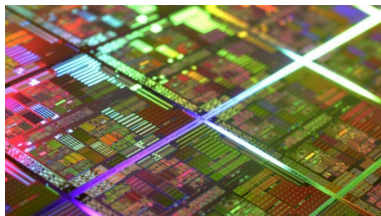
Constrained Deformation for Evolutionary Optimization

Daniel Sieger

Graphics & Geometry Group, Bielefeld University

Computer-aided Design and Engineering

Tools and practices for creating highly complex industrial products



Design Optimization and Shape Deformation

Design optimization

- Discover alternative designs with improved properties
- Decrease the need for costly real-world prototypes

Shape deformation

- Effective means to create design variations during optimization

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Initial Design



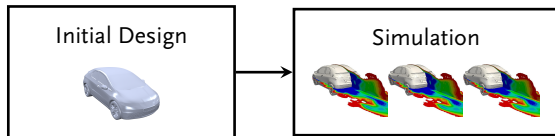
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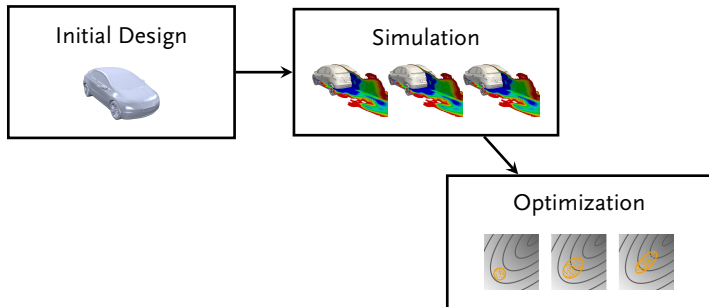
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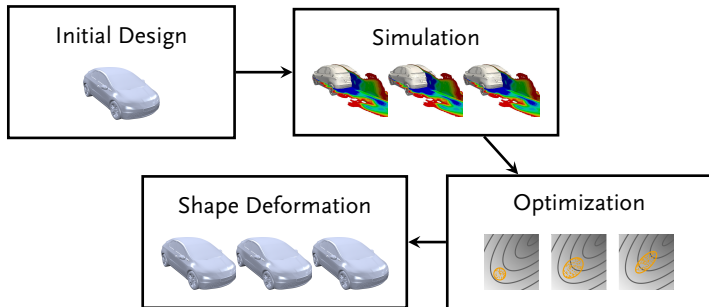
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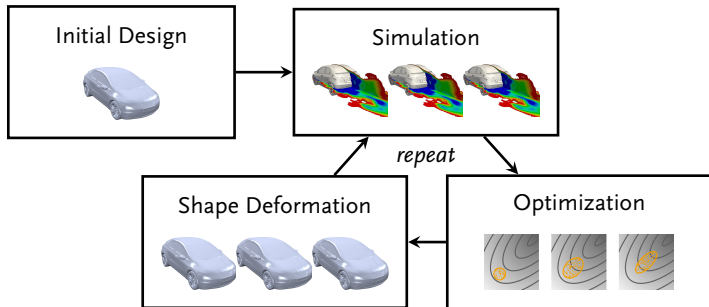
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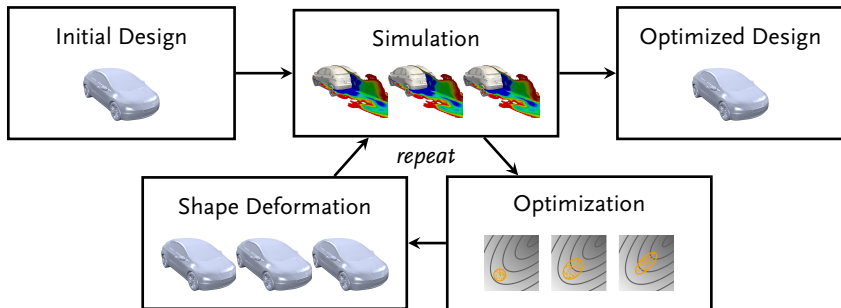
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Key Questions

1. What is a *good* deformation technique for design optimization?
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 - Effective application within design optimization
 - Address specific challenges

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 - Fundamental requirements and properties
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2. How to *apply* and *improve* existing techniques?
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3. How to incorporate *geometric constraints* into the deformation?
 - Maintain critical properties such as planarity
 - Maintain deformation quality and flexibility

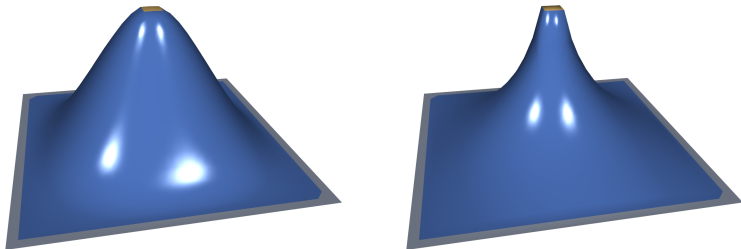
Outline

1. Shape deformation methods for design optimization
 - Introduce the state of the art
 - Benchmarks and comparisons
2. Advanced RBF deformation techniques
 - Automatic CAD-based design optimization
 - Avoid costly remeshing
3. Constrained deformation
 - Effective constraint preservation
 - High-quality, flexible deformations

Shape Deformation for Design Optimization

Thin Shell Deformation ¹

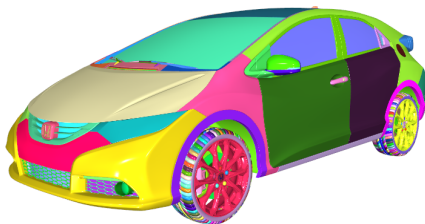
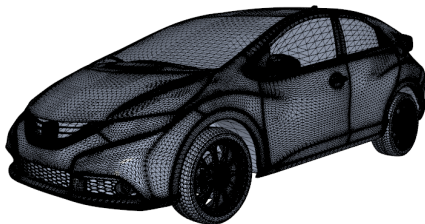
- Physically-inspired technique suitable for sheet metal surfaces
- Flexible modeling of material behavior
- Based on the minimization of stretching and bending energies



1. Botsch et al., *On Linear Variational Surface Deformation Methods*, Trans. on Visualization and Computer Graphics, 2008

Limitations

Assumption: Surface $\mathcal{S}$ is a proper triangle mesh



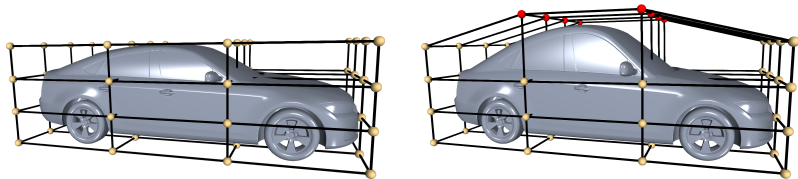
Space Deformation Methods

- Instead of deforming the surface $\mathcal{S}$, deform embedding space Ω
 - + Representation independence
 - + Disconnected components
 - + Robustness against defects
- Construct a space deformation function $\boldsymbol{d}: \Omega \subset \mathbb{R}^3 \rightarrow \mathbb{R}^3$

Free-Form Deformation² (FFD)

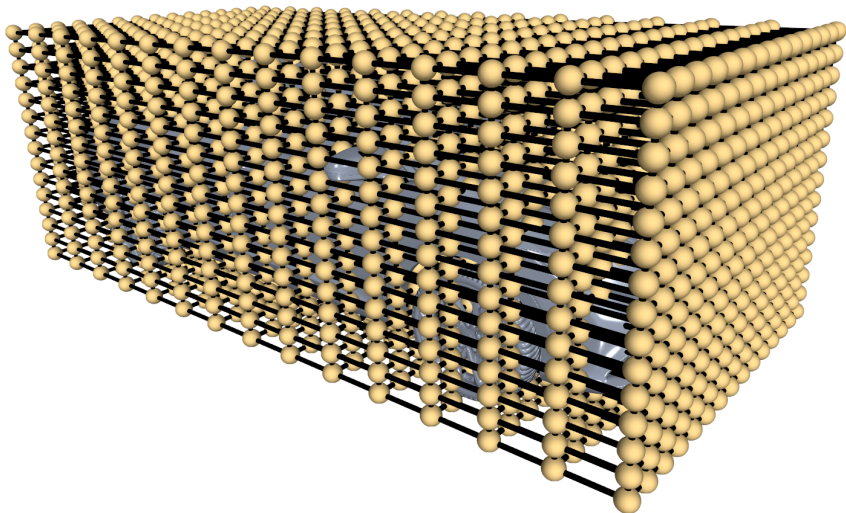
- Idea: Embed object in control grid and deform grid
- Procedure:
 1. Compute local coordinates within the control grid
 2. Select and move control points
 3. Deform object according to updated control points

$$\mathbf{d}(\mathbf{x}) = \sum_{i=0}^l \sum_{j=0}^m \sum_{k=0}^n \delta \mathbf{c}_{ijk} \varphi_i(u_1) \varphi_j(u_2) \varphi_k(u_3),$$



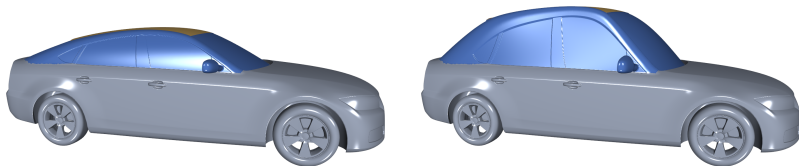
FFD Limitations

Manipulating large control grids by hand is tedious



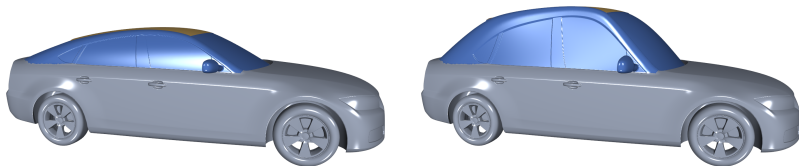
Direct Manipulation FFD³

- Move object points directly (constraints)
- Compute control point displacements satisfying constraints
- Requires solving a linear system (pseudo-inverse, SVD)



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- Move object points directly (constraints)
- Compute control point displacements satisfying constraints
- Requires solving a linear system (pseudo-inverse, SVD)
- Minimizes constraint error and control point movement
- Not physically plausible, not precise enough



Radial Basis Function Deformation⁴

- Deformation as scattered data interpolation problem
 - *Exactly* interpolate prescribed displacements
 - *Smoothly* interpolate displacements through space

4. Botsch and Kobbelt, *Real-Time Shape Editing Using Radial Basis Functions*, Computer Graphics Forum, 2005

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$$\mathbf{d}: \mathbb{R}^3 \rightarrow \mathbb{R}^3$$

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$$\mathbf{d}(\mathbf{x}) = \sum_{j=1}^m \mathbf{w}_j \varphi_j(\mathbf{x}) + \boldsymbol{\pi}(\mathbf{x})$$

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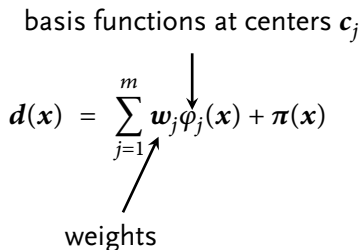

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weights polynomial term

A diagram showing the equation $\mathbf{d}(\mathbf{x}) = \sum_{j=1}^m \mathbf{w}_j \phi_j(\mathbf{x}) + \boldsymbol{\pi}(\mathbf{x})$. Above the equation, the text "basis functions at centers $\mathbf{c}_j$ " has a downward arrow pointing to $\phi_j(\mathbf{x})$. Below the equation, the text "weights" has an upward arrow pointing to $\mathbf{w}_j$, and the text "polynomial term" has an upward arrow pointing to $\boldsymbol{\pi}(\mathbf{x})$.

Radial Basis Functions

- Various choices: Gaussian, multiquadrics, thin plate spline...
- Choose triharmonic basis functions $\varphi(r) = r^3$ so that $\mathbf{d}$ minimizes fairness energy:

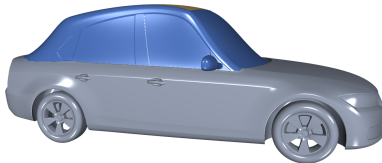
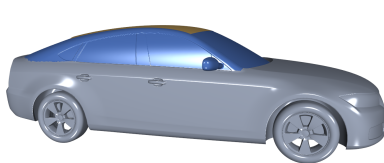
$$\int_{\mathbb{R}^3} \left\| \frac{\partial^3 \mathbf{d}}{\partial x^3} \right\|^2 + \left\| \frac{\partial^3 \mathbf{d}}{\partial x^2 \partial y} \right\|^2 + \dots + \left\| \frac{\partial^3 \mathbf{d}}{\partial z^3} \right\|^2 dV$$

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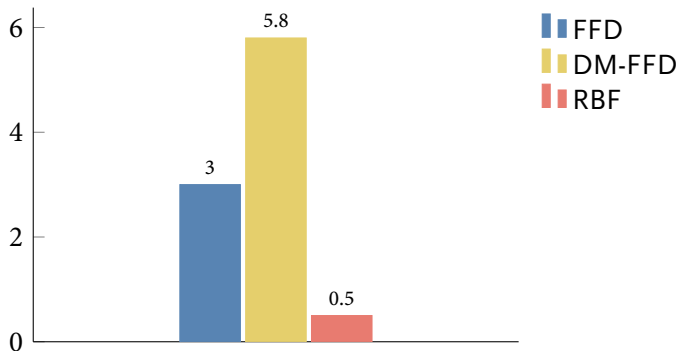
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- Determine weights and polynomial coefficients
→ Solve linear system



Performance Comparison⁵

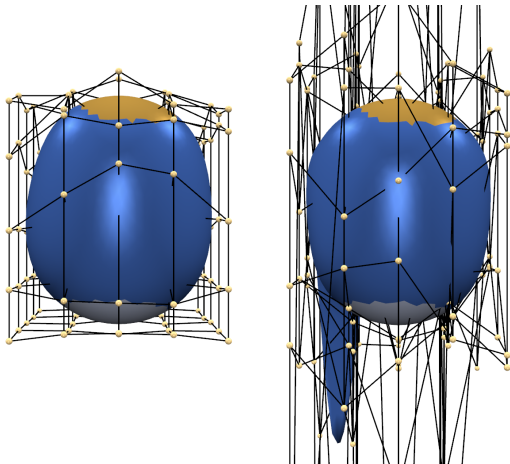
Precomputed deformation, times in seconds



5. Sieger et al., *On Shape Deformation Techniques for Simulation-based Design Optimization*, New Challenges in Grid Generation and Adaptivity for Scientific Computing, 2015

Robustness

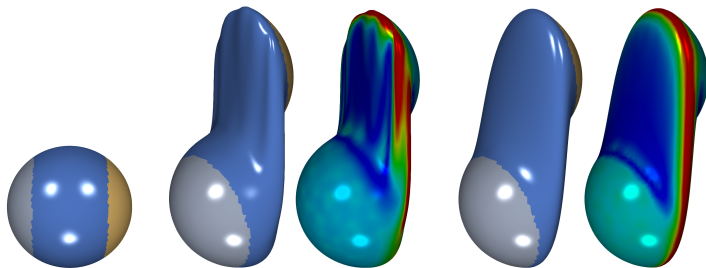
DM-FFD: Solving the linear system using SVD requires clamping of small singular values. Since the clamping value is not known in advance, robustness artifacts might occur.



Quality

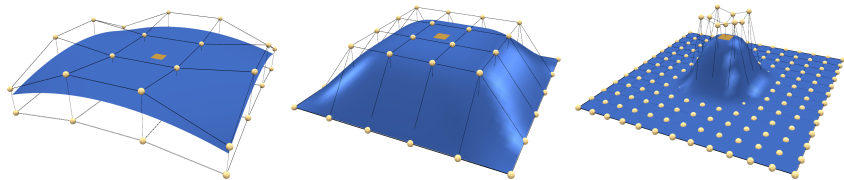
Quality comparison using mean curvature:

- FFD methods (center) might lead to smoothness artifacts
- RBFs (right) yield highly smooth results



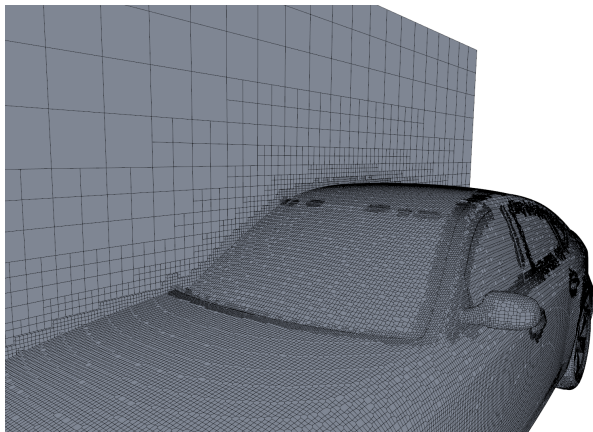
Quality

FFD: The shape of the deformation depends on grid resolution and not on some form of physically-inspired energy minimization



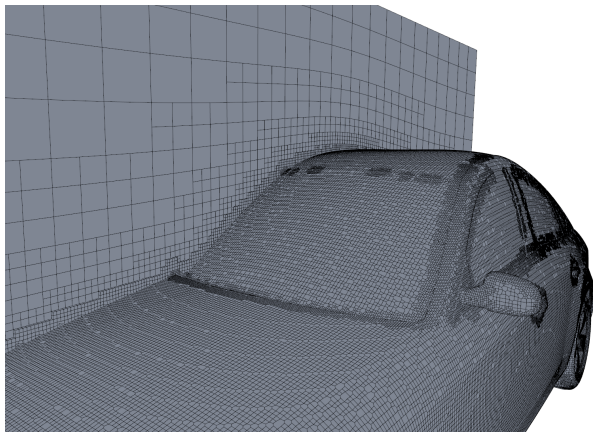
Volume Mesh Deformation

- Deformation of a hex-dominant volume mesh for CFD simulation



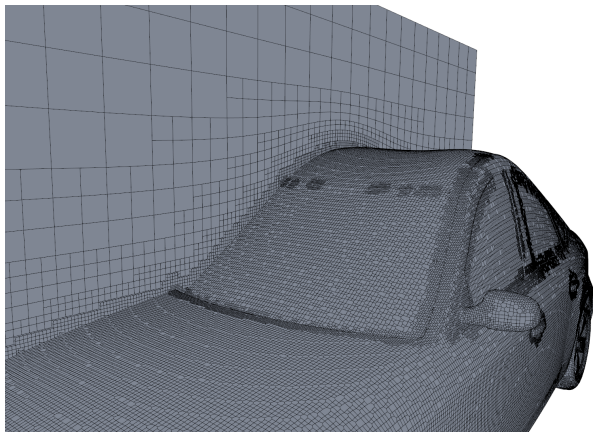
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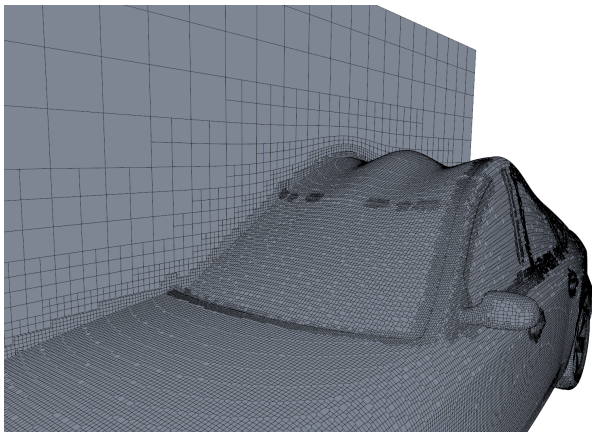
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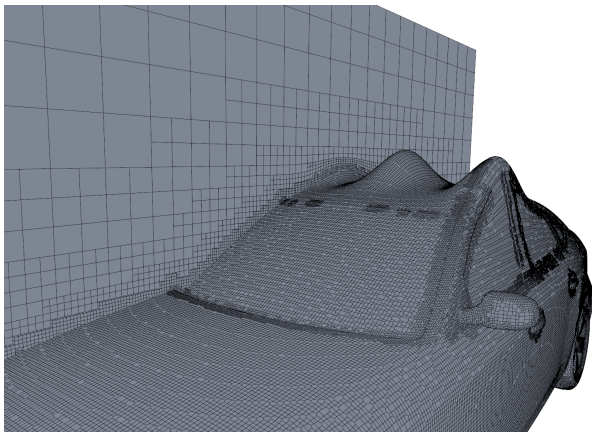
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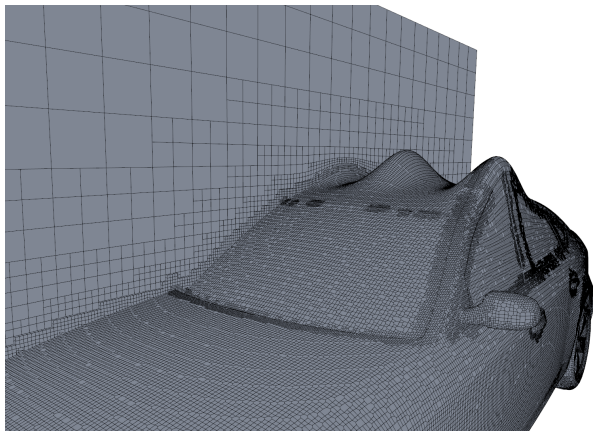
Volume Mesh Deformation

- Deformation of a hex-dominant volume mesh for CFD simulation



Volume Mesh Deformation

- Deformation of a hex-dominant volume mesh for CFD simulation
- Depending on the FFD grid resolution the mesh becomes unusable for simulation



Evaluation Summary

	Performance	Robustness	Quality	Adaptivity	Precision
FFD	•	+	•	-	-
DM-FFD	-	•	•	-	•
RBF	•	+	+	+	+

• : neutral + : positive - : negative

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Advanced RBF Deformation Techniques

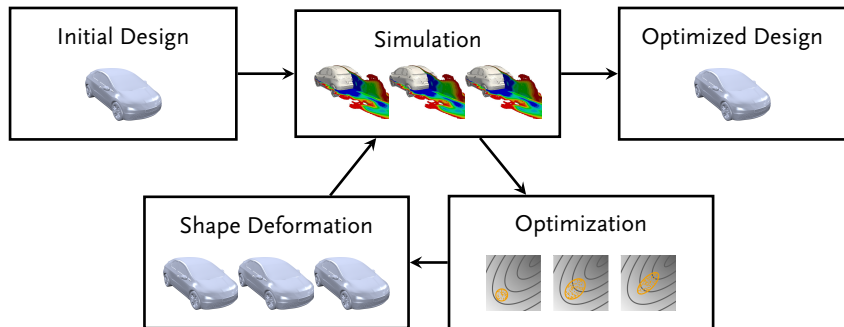
Advanced RBF Deformation Techniques ⁶

- Unified framework for combined surface and volume deformation
- Preventing self-intersections by splitting
- Performance improvements through advanced linear solvers

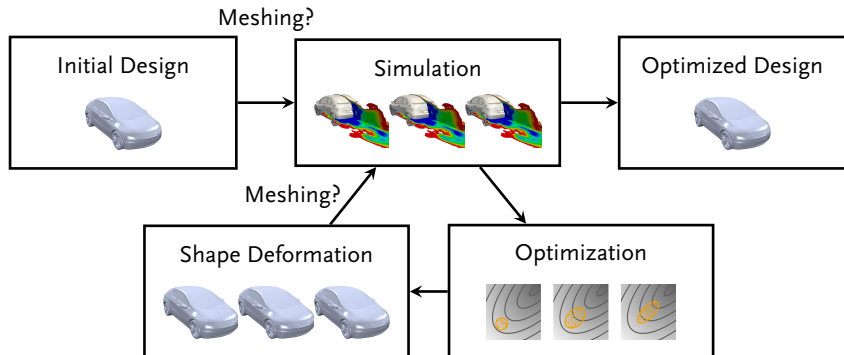
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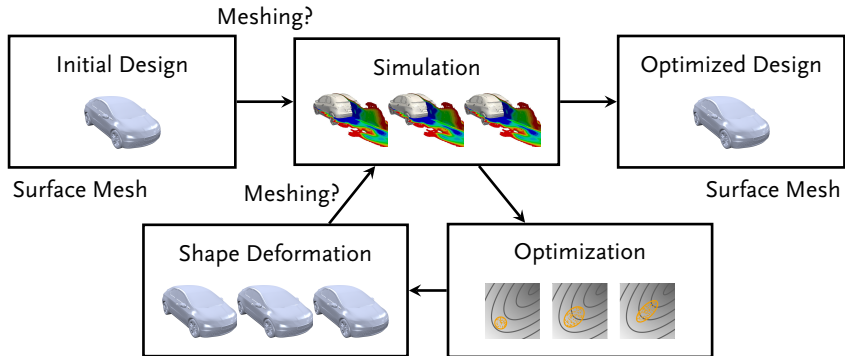
Shape Deformation & Design Optimization



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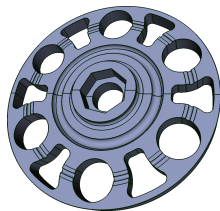


Shape Deformation & Design Optimization

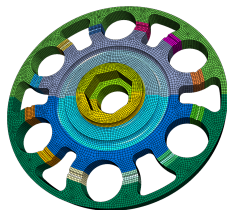
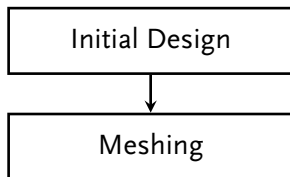


Generalized Design Optimization Scenario

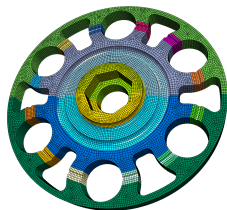
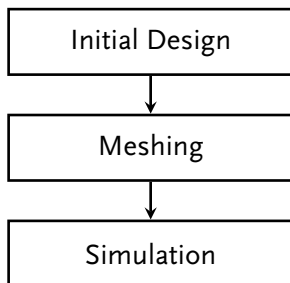
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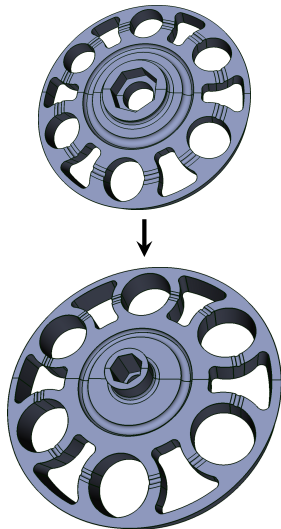
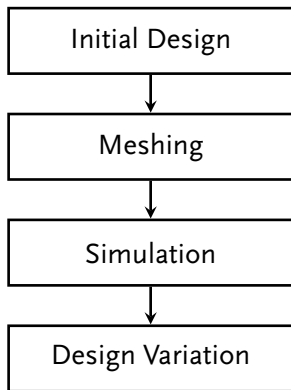
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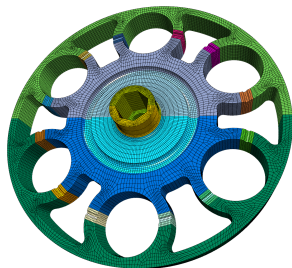
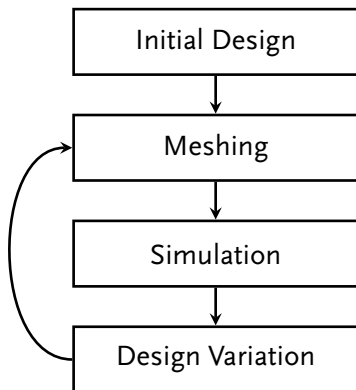
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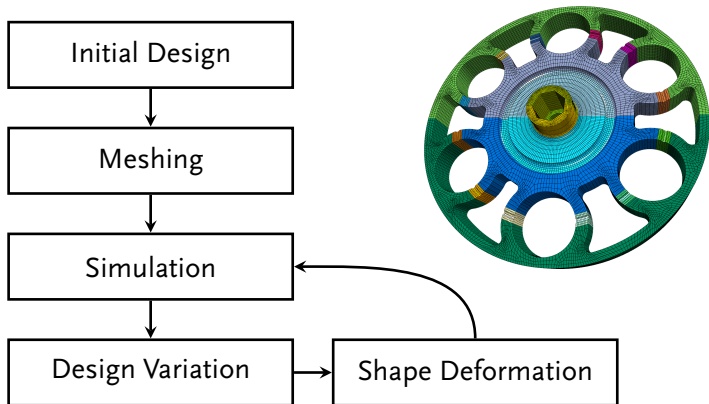
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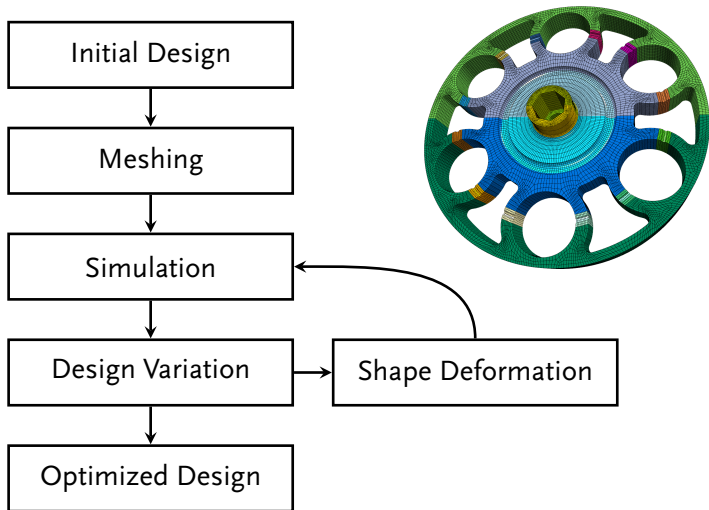
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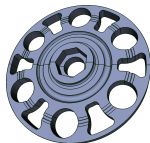
Unified Framework

Procedure:

1. Update CAD model
2. Surface deformation:
Compute surface points matching updated CAD model
3. Volume deformation:
Compute volume points matching updated surface points

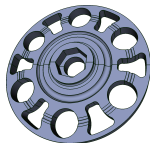
Surface Deformation

CAD

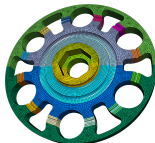


Surface Deformation

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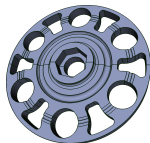


Mesh

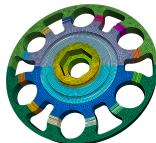


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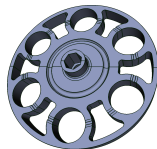
CAD



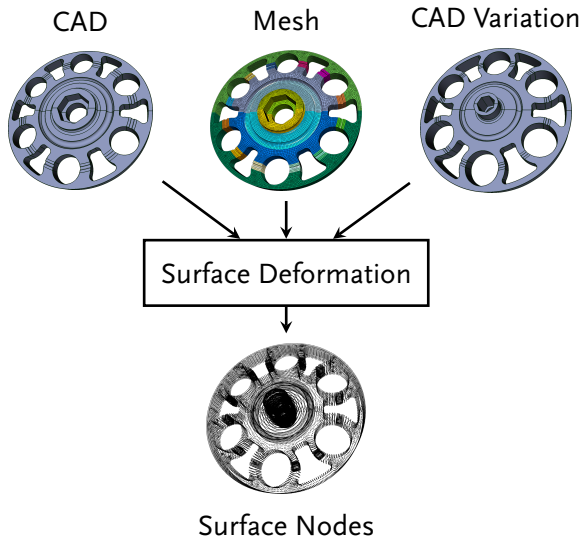
Mesh



CAD Variation

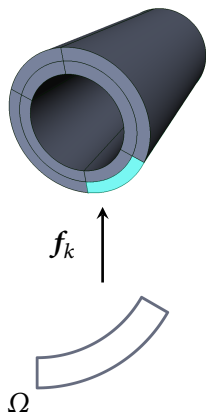


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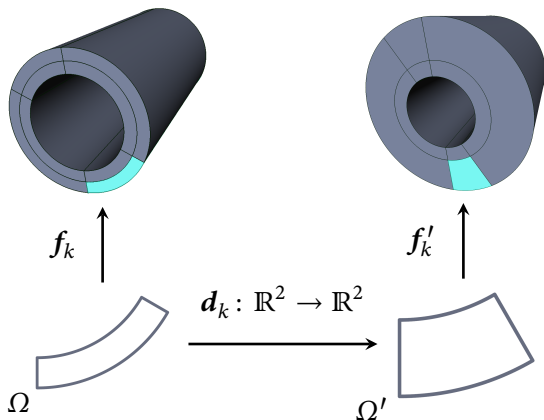
Surface Deformation

1. Determine parametric coordinates for each surface node

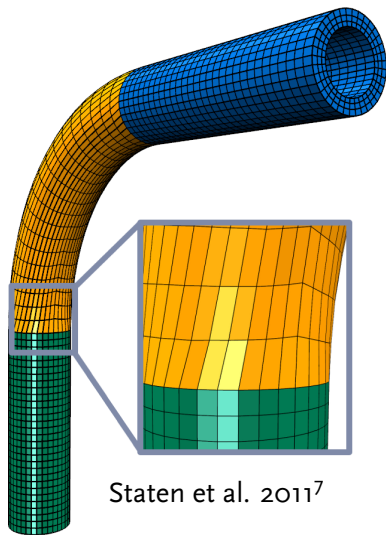


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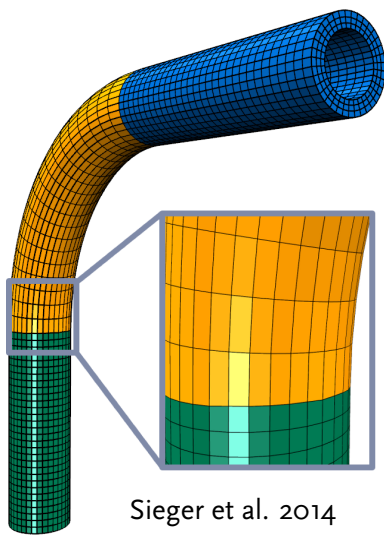
1. Determine parametric coordinates for each surface node
2. Construct 2D RBF deformation to warp coordinates



Surface Deformation



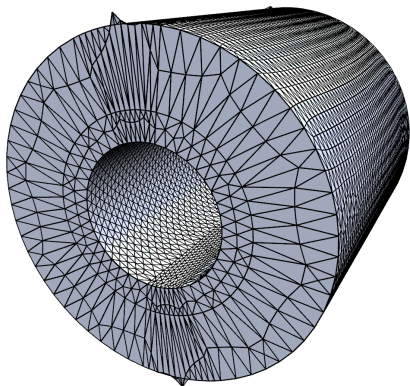
Staten et al. 2011⁷



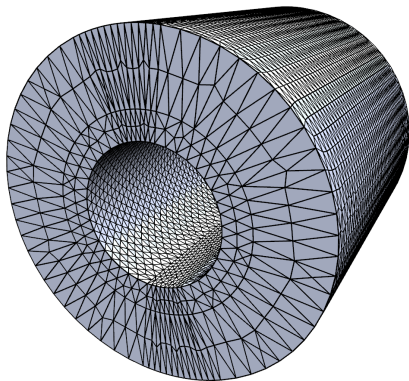
Sieger et al. 2014

7. Staten et al., *A Comparison of Mesh Morphing Methods for 3D Shape Optimization*, IMR 2011

Surface Deformation

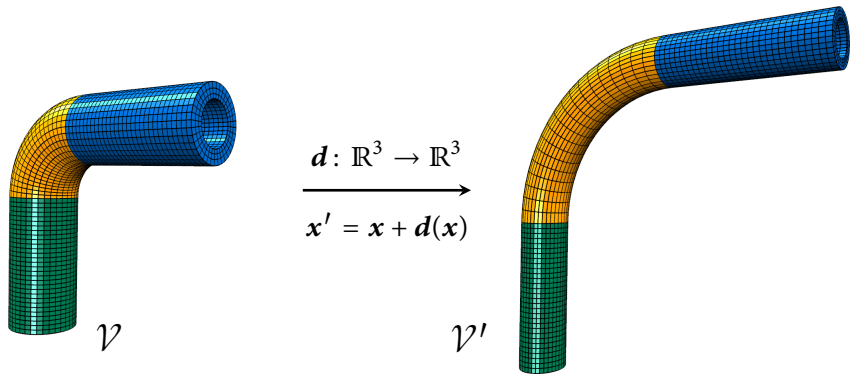


Staten et al. 2011



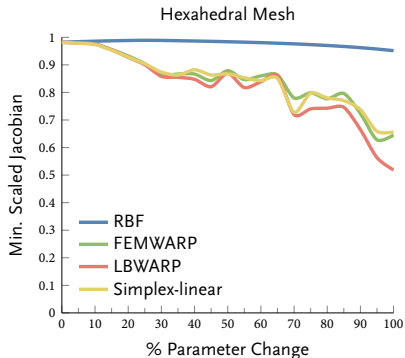
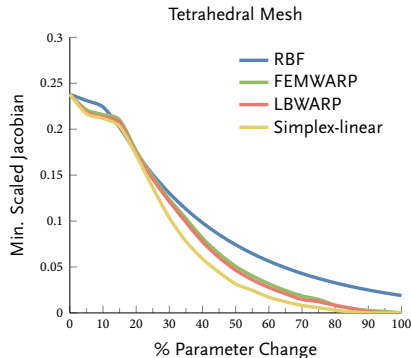
Sieger et al. 2014

Volume Deformation



Volume Deformation Quality

- Comparison with best techniques from Staten et al. 2011
- Min. scaled Jacobian vs. parameter change in the CAD model
- Tetrahedral and hexahedral volume meshes



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By *exploiting the flexibility of RBFs*, we can implement fully automatic design optimization loops based on CAD prototypes.

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Constrained Deformation

Constrained Space Deformation for Design Optimization⁷

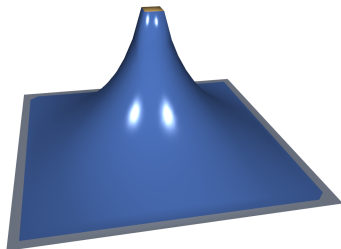
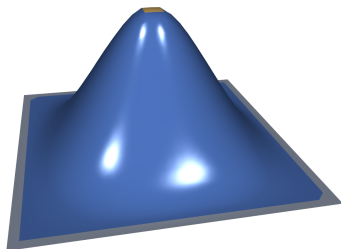
Combines

- Modeling flexibility of surface-based methods
- Representation-independence of space deformations
- Quality of RBFs
- Improved scalability
- Geometric constraints

7. Sieger et al., *Constrained Space Deformation for Design Optimization*, Computer-Aided Design 2016

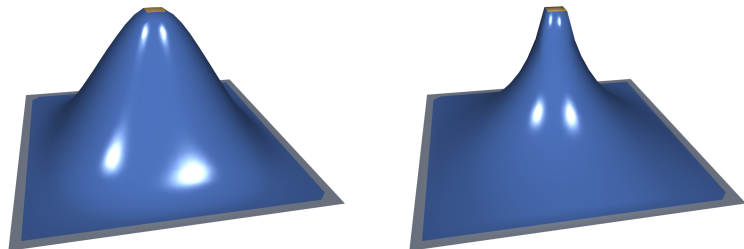
Surface Deformation Revisited

- Surface-based methods allow for flexible modeling of material behavior by explicitly choosing the deformation energy to be minimized



Surface Deformation Revisited

- Surface-based methods allow for flexible modeling of material behavior by explicitly choosing the deformation energy to be minimized
- Goal: Achieve the same flexibility using space deformations



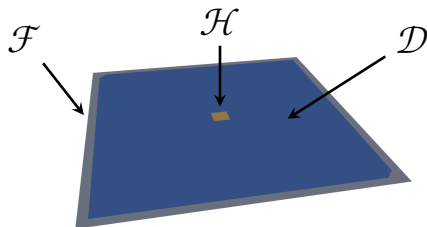
Thin Shell Deformation

Measure stretching and bending by 1st and 2nd order partial derivatives of the displacement function $\mathbf{d}: S \rightarrow \mathbb{R}^3$

$$E_{\text{stretch}}[\mathbf{d}] = \int_{\mathcal{D}} \|\nabla \mathbf{d}(\mathbf{x})\|^2 d\mathbf{x}$$

$$E_{\text{bend}}[\mathbf{d}] = \int_{\mathcal{D}} \|\Delta \mathbf{d}(\mathbf{x})\|^2 d\mathbf{x}$$

$$E_{\text{fix}}[\mathbf{d}] = \int_{\mathcal{H} \cup \mathcal{F}} \|\mathbf{d}(\mathbf{x}) - \bar{\mathbf{d}}(\mathbf{x})\|^2 d\mathbf{x}$$



Surface-Based Discretization

- Discretize on the mesh using standard differential operators ⁸

$$E_{\text{stretch}}[\mathbf{d}] = \sum_{t \in \mathcal{D}} A_t \|\nabla \boldsymbol{\delta}_t\|^2$$

$$E_{\text{bend}}[\mathbf{d}] = \sum_{\mathbf{x}_i \in \mathcal{D}} A_i \|\Delta \boldsymbol{\delta}_i\|^2$$

$$E_{\text{fix}}[\mathbf{d}] = \sum_{\mathbf{x}_i \in \mathcal{H} \cup \mathcal{F}} A_i \|\boldsymbol{\delta}_i - \bar{\boldsymbol{\delta}}_i\|^2$$

→ Solve a linear system:

$$\left[\gamma_s \mathbf{G}^T \mathbf{G} + \gamma_b \mathbf{L}^T \mathbf{L} + \gamma_f \mathbf{F}^T \mathbf{F} \right] \mathbf{D} = \gamma_f \mathbf{F}^T \mathbf{F} \bar{\mathbf{D}}$$

8. Meyer et al., *Discrete Differential-Geometry Operators for Triangulated 2-Manifolds*, Visualization and Mathematics, 2003

Space Deformation Methods

- Construct a space deformation function $\boldsymbol{d}: \Omega \subset \mathbb{R}^3 \rightarrow \mathbb{R}^3$
 - Use meshless approximation methods
 - Represent $\boldsymbol{d}$ by basis functions φ_j located at centers $\boldsymbol{c}_j$:

$$\boldsymbol{d}(\boldsymbol{x}) = \sum_{j=1}^k \boldsymbol{w}_j \varphi_j(\boldsymbol{x})$$

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 - Which basis functions to choose?
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Surface Sampling

Goal: Generate uniformly distributed points $\mathbf{c}_j$ on the surface

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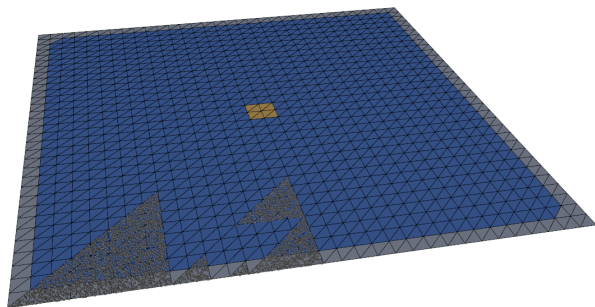
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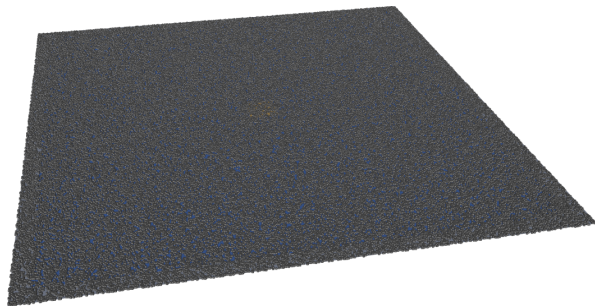
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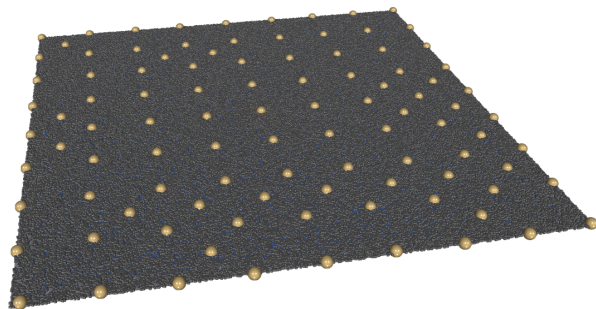
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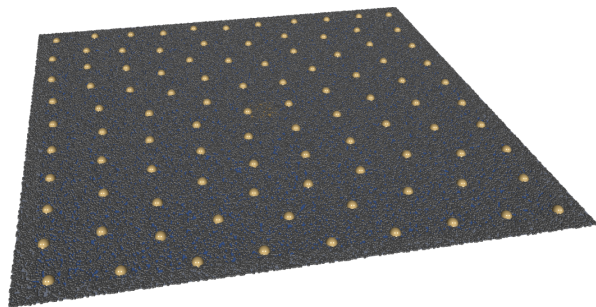
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Surface Sampling

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2. Farthest point selection
3. Lloyd relaxation (k-means clustering)



Subspace Surface Deformation

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$$\mathbf{D} = \Phi \mathbf{W} \quad \text{with} \quad \Phi_{ij} = \varphi_j(\mathbf{x}_i)$$

- Leads to a modified linear system:

$$\Phi^T [\gamma_s \mathbf{G}^T \mathbf{G} + \gamma_b \mathbf{L}^T \mathbf{L} + \gamma_f \mathbf{F}^T \mathbf{F}] \Phi \mathbf{W} = \Phi^T [\gamma_f \mathbf{F}^T \mathbf{F} \bar{\mathbf{D}}]$$

Subspace Surface Deformation

- Global triharmonic RBFs

$$\varphi_j(\mathbf{x}) = \|\mathbf{x} - \mathbf{c}_j\|^3$$

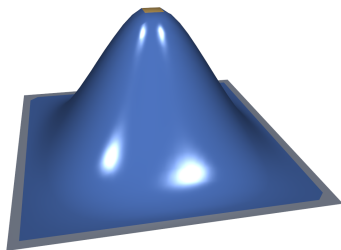
- Energy minimization built-in

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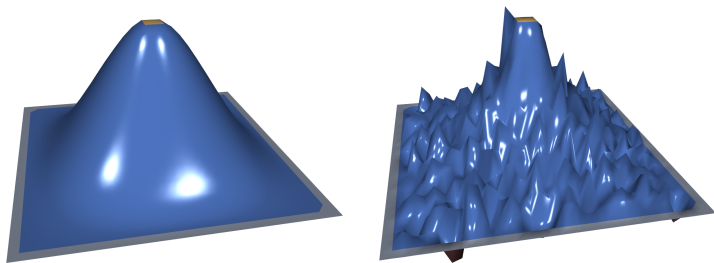


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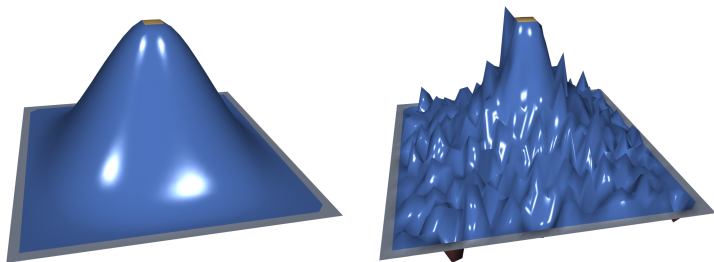


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- Global triharmonic RBFs

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- Good results for bending, bad results for stretching
- Global support $\rightarrow$ dense linear systems



Subspace Surface Deformation

- Compactly supported RBFs

$$\varphi_j(\mathbf{x}) = \varphi(\|\mathbf{x} - \mathbf{c}_j\|) = \varphi(r) = \begin{cases} (1-r)^4(4r+1), & r < \sigma, \\ 0, & \text{otherwise.} \end{cases}$$

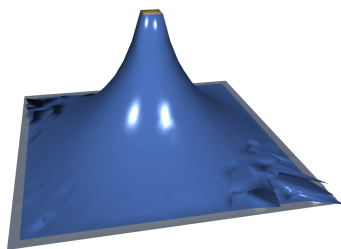
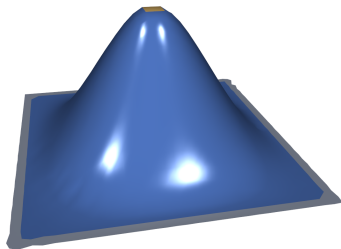
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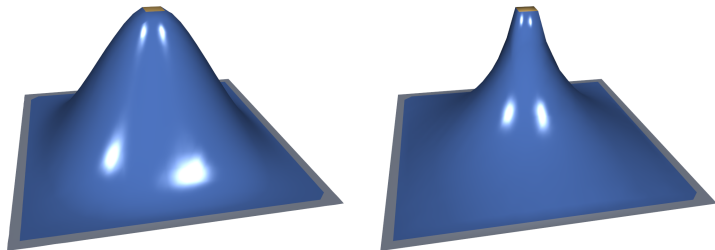


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Subspace Surface Deformation

- Moving Least Squares (MLS) basis functions⁹

$$\varphi_j(\mathbf{x}) = \mathbf{p}(\mathbf{x})^T \mathbf{M}^{-1}(\mathbf{x}) \mathbf{p}(\mathbf{c}_j) w(\mathbf{x} - \mathbf{c}_j)$$

9. Fries et al., *Classification and Overview of Meshfree Methods*, Technical Report, TU Braunschweig, 2003

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polynomial basis



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Diagram illustrating the components of the Moving Least Squares (MLS) basis function $\varphi_j(\mathbf{x})$:

- The term $\mathbf{p}(\mathbf{x})^T$ is labeled as the **polynomial basis** (indicated by an arrow).
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- $\mathbf{M}^{-1}(\mathbf{x})$ is labeled as the **4 × 4 moment matrix**.
- $\mathbf{p}(\mathbf{c}_j)$ and $w(\mathbf{x} - \mathbf{c}_j)$ are collectively labeled as the **weighting function**.

Subspace Surface Deformation

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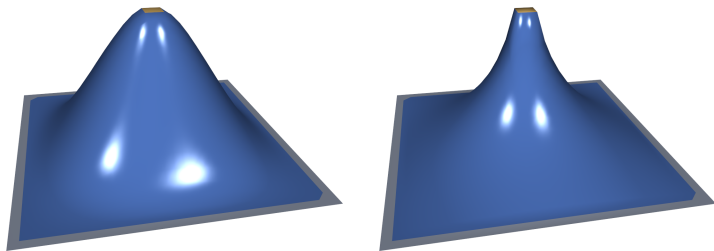
$$\varphi_j(\mathbf{x}) = \underset{\substack{\uparrow \\ \text{polynomial basis}}}{\mathbf{p}(\mathbf{x})}^{\text{T}} \overset{\substack{\downarrow \\ 4 \times 4 \text{ moment matrix}}}{\mathbf{M}^{-1}(\mathbf{x})} \underset{\substack{\nwarrow \\ \text{weighting function}}}{\mathbf{p}(\mathbf{c}_j)} w(\mathbf{x} - \mathbf{c}_j)$$

- More complex form, inversion of $\mathbf{M}$ required

9. Fries et al., *Classification and Overview of Meshfree Methods*, Technical Report, TU Braunschweig, 2003

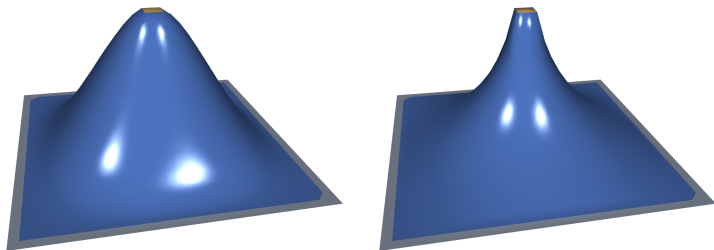
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Volumetric Space Deformation

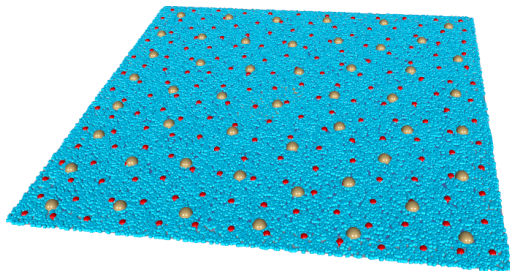
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- Replace vertex-based integration over the surface $\mathcal{S}$ with purely space-based integration method
- Use Lloyd-based sampling to determine integration points $\mathbf{t}_i$



Volumetric Space Deformation

- Evaluate gradients and Laplacians of φ_j at integration points $\mathbf{t}_i$:

$$E_{\text{stretch}} = \sum_{i=1}^N V_i \|\nabla \mathbf{d}(\mathbf{t}_i)\|^2 = \sum_{i=1}^N V_i \left\| \sum_{j=1}^k \mathbf{w}_j \nabla \varphi_j(\mathbf{t}_i) \right\|^2$$

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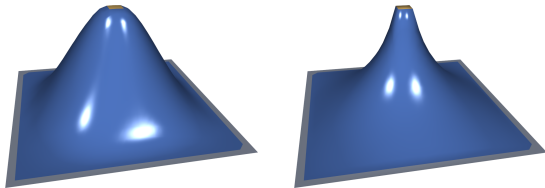
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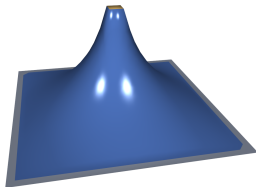
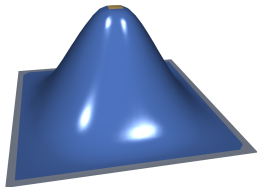
Volumetric Space Deformation

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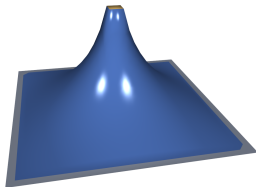
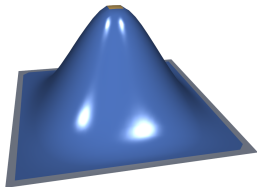


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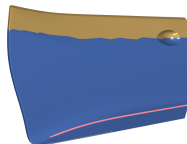
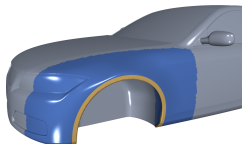
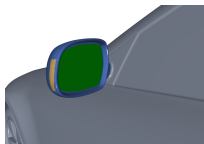


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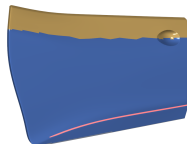
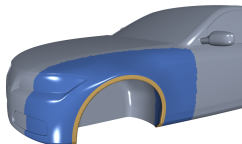
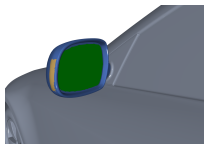
Geometric Constraints

- Design prototypes contain important geometric features
 - Planar components
 - Circular couplings or wheelhouses
 - Characteristic feature lines



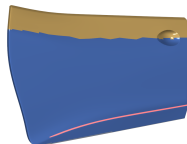
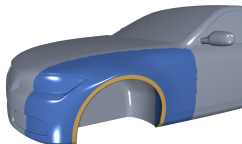
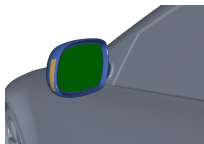
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- Deforming the design during optimization distorts features
 - Impaired functionality
 - Violated production limitations
- Classical solution: Add penalty terms to the cost function
 - *Creation* of infeasible designs
 - Costly *evaluation* (e.g., CFD)



Geometric Constraints

- Approach: Prevent distortion of features by incorporating geometric constraints into the deformation
 - + Only create feasible designs
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 - Most methods are surface-based
 - + Projection-based constraints¹⁰

10. Bouaziz et al., *Shape-Up: Shaping Discrete Geometry with Projections*, Computer Graphics Forum, 2012

Projection-Based Constraints

- Define projection operators P_t : Plane, circle, ...
- Measure deviation from prescribed constraints

$$E_{\text{constr}}(\mathbf{X}) = \sum_{t=1}^m \|\mathbf{X} - P_t(\mathbf{X})\|^2$$

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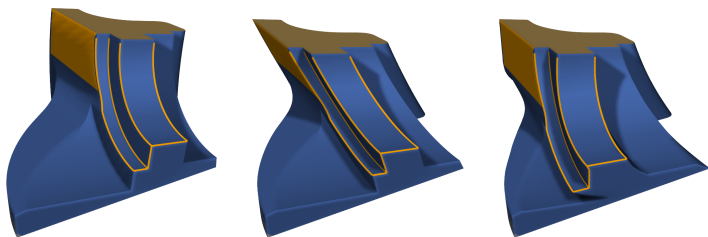
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- Minimize E_{constr} by iterative alternating optimization
- Integrated into deformation framework
 - Extended original formulation to scale to large constraint areas

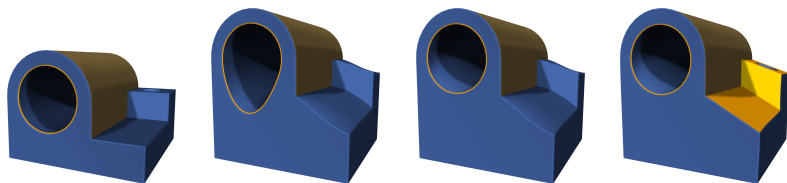
Geometric Constraint Examples

Fundamental geometric constraints: Planarity, circularity, feature lines



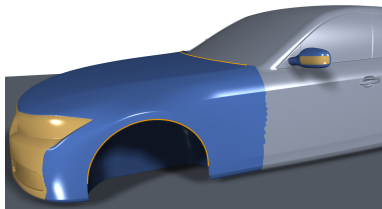
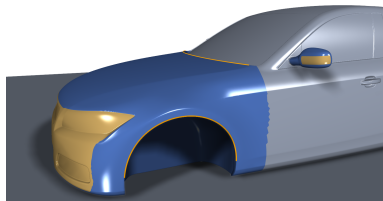
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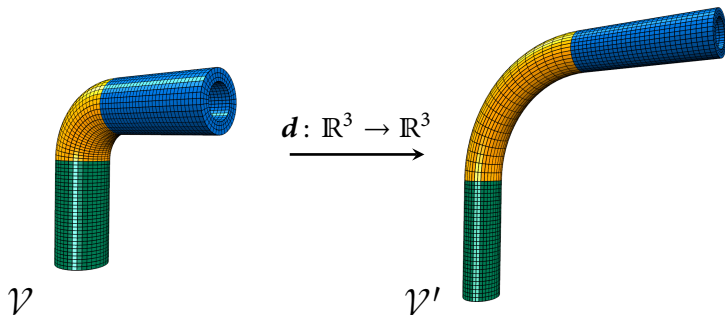
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Volume Deformation Examples

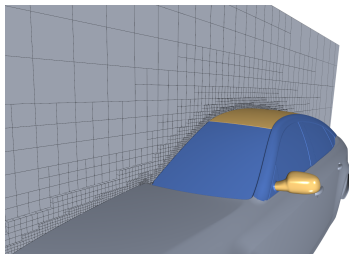
Comparison to previous results¹¹

- Original mesh quality: 0.98
- After RBF deformation: 0.951
- Using our new method: 0.954

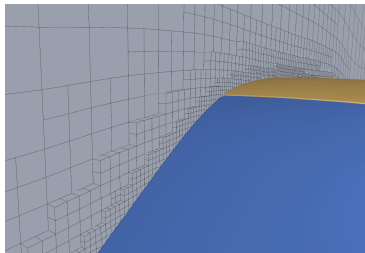
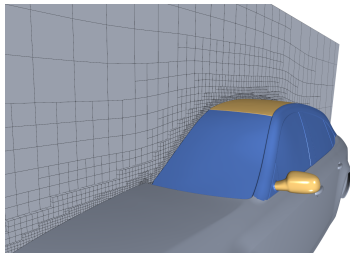
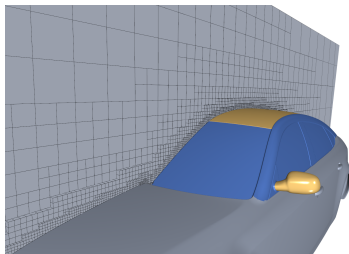


¹¹. Sieger et al., *RBF Morphing Techniques for Simulation-based Design Optimization*, Engineering with Computers, 2014

Combined Surface and Volume Deformation



Combined Surface and Volume Deformation



Constrained Deformation Summary

- A deformation technique for design optimization
 - + Modeling flexibility like surface-based methods
 - + Representation-independence of space deformations
 - + High quality comparable to RBFs
 - + Improved scalability through sparse linear systems
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- Observation: Significant differences in the modeling flexibility and quality of RBFs and MLS

Directions for Future Work

- Additional constraint types, semantic relations
 - Mutual distances
 - Symmetry
 - Co-planarity
- More tight integration with CAD tools
 - Integration into CAD-based optimization framework
 - Transfer of constraints and functional requirements
- More extensive evaluation

Key Questions and Answers

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The combination of *MLS space deformations and projective constraints* allows for the creation of more feasible designs.

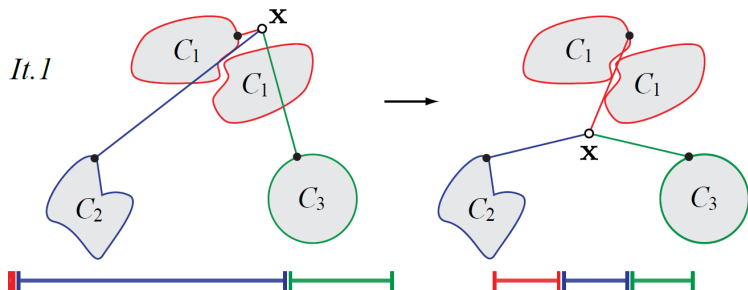
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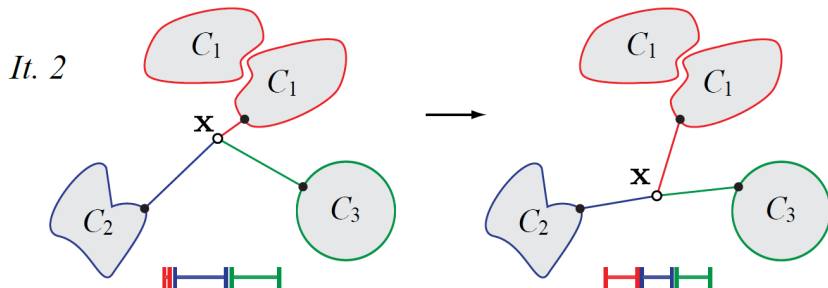
Projection-Based Constraints

- Alternating optimization:
 1. Project current point on constraint
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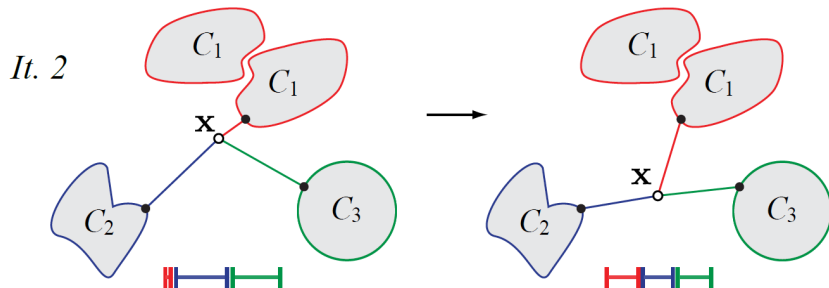
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- + Flexible and simple formulation
- Slow convergence for complex constraints

