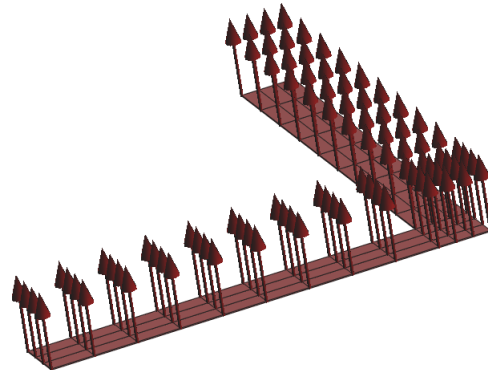




University of Stuttgart
Institute for Structural Mechanics

Application of optimization on manifolds to nonlinear shell theory

Alexander Müller, Manfred Bischoff

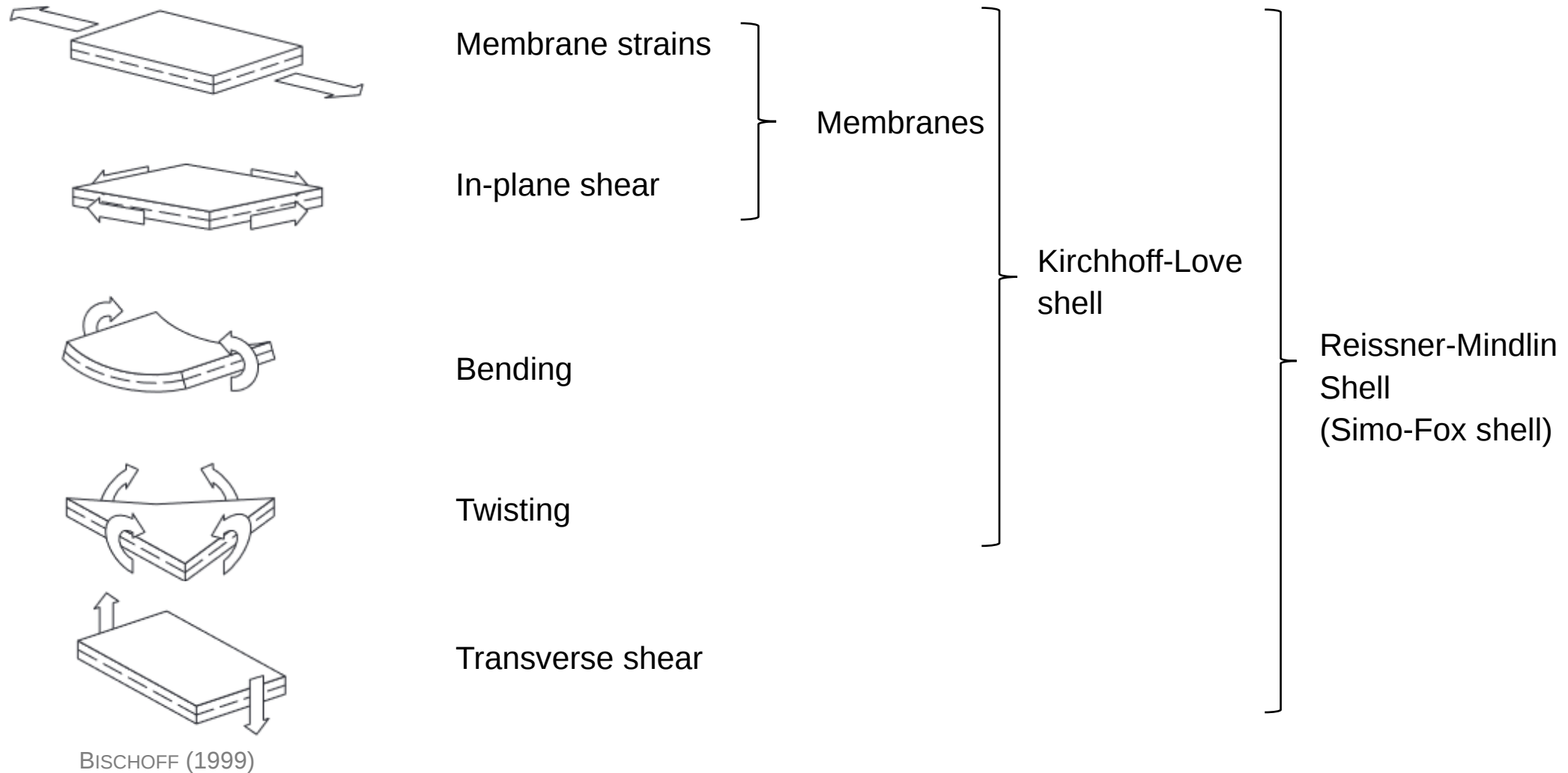


**Workshop on
Nonlinear Bending,
University of Freiburg**

24.05.2022

Kinematics of a nonlinear Reissner-Mindlin shell model

Energy contributions



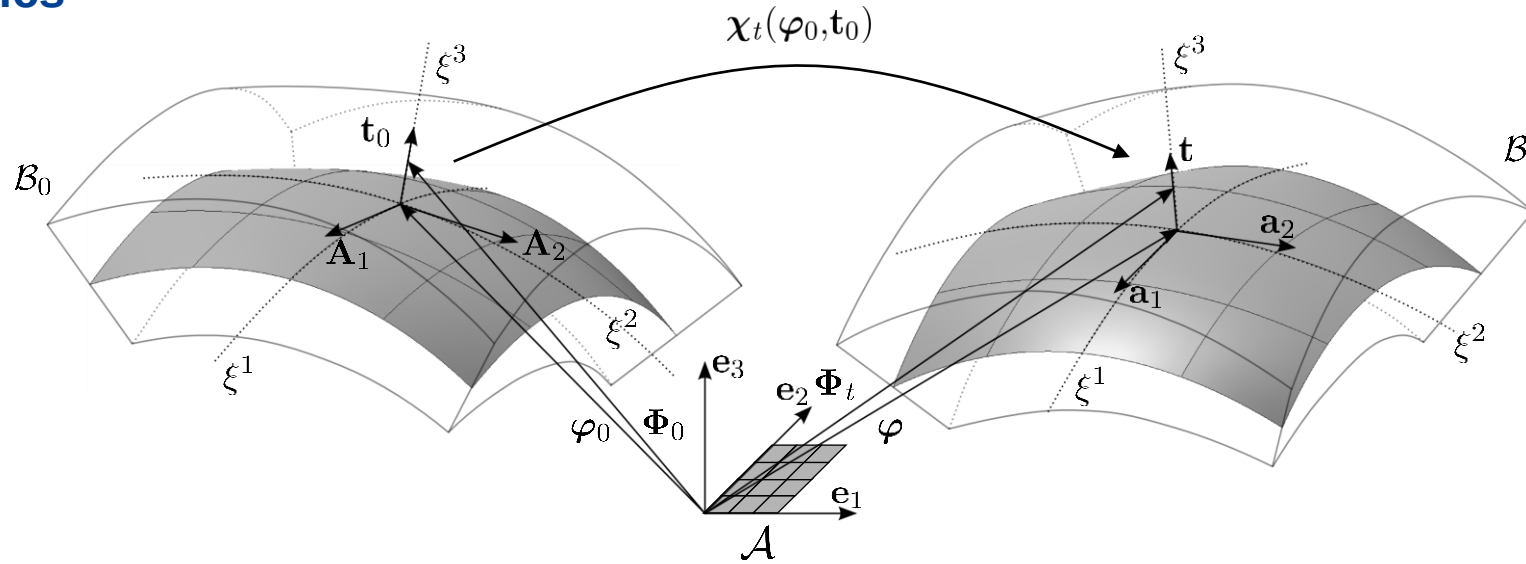
Outline

Brief review of the kinematics of a nonlinear Reissner-Mindlin shell model

Challenges for a consistent formulation

Overcoming these challenges with optimization on manifolds

Kinematics



$$\boldsymbol{\xi} = [\xi^1, \xi^2, \xi^3]^T \in \mathcal{A}$$

Mappings:

$$\Phi_0 : \begin{cases} \mathcal{A} \rightarrow \mathcal{B}_0 \subset \mathbb{R}^3 \\ \boldsymbol{\xi} \mapsto \mathbf{X} = \Phi_0(\boldsymbol{\xi}) \end{cases}$$

$$\begin{aligned} \mathbf{X} &= \Phi_0(\xi^1, \xi^2, \xi^3) = \boldsymbol{\varphi}_0(\xi^1, \xi^2) + \xi^3 \mathbf{t}_0(\xi^1, \xi^2) \\ \mathbf{x} &= \Phi_t(\xi^1, \xi^2, \xi^3) = \boldsymbol{\varphi}(\xi^1, \xi^2) + \xi^3 \mathbf{t}(\xi^1, \xi^2) \end{aligned}$$

$$\Phi_t : \begin{cases} \mathcal{A} \rightarrow \mathcal{B}_t \subset \mathbb{R}^3 \\ \boldsymbol{\xi} \mapsto \mathbf{x} = \Phi_t(\boldsymbol{\xi}) \end{cases}$$

$$\chi_t = \Phi_t \circ \Phi_0^{-1}, \quad \chi_t : \begin{cases} \mathcal{B}_0 \rightarrow \mathcal{B}_t \subset \mathbb{R}^3 \\ \mathbf{X} \mapsto \mathbf{x} = \chi_t(\mathbf{X}) \end{cases}$$

Kinematic assumptions

- Geometry approximation:

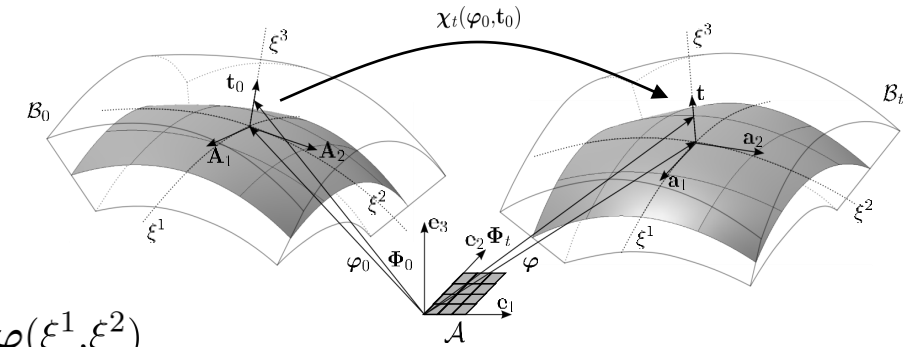
$$\mathbf{x} = \Phi_t(\xi^1, \xi^2, \xi^3) = \varphi(\xi^1, \xi^2) + \xi^3 \mathbf{t}(\xi^1, \xi^2)$$

- First order transverse shear effects are taken into account:

- The director field $\mathbf{t}(\xi^1, \xi^2)$ is independent of the midsurface field $\varphi(\xi^1, \xi^2)$ (in contrast to Kirchhoff-Love)

- Thickness change is not contained in kinematic description:

- The director is a unit vector, i.e. $\mathbf{t} : \mathcal{A} \rightarrow \mathcal{S}^2 \subset \mathbb{R}^3$ $\mathcal{S}^2 = \{\mathbf{x} \in \mathbb{R}^3 \mid \mathbf{x} \cdot \mathbf{x} = 1\}$



Discretization

- Discretization of the midsurface field $\varphi : \mathcal{A} \rightarrow \mathbb{R}^3$ trivial since it maps onto a vector space $\mathbb{R}^3$
- Discretization and parametrization of the director field $\mathbf{t} : \mathcal{A} \rightarrow \mathcal{S}^2$ difficult due to the nonlinear space $\mathcal{S}^2$

Challenges

Interpolation on the unit sphere

(Algebraic) optimization on manifolds

Update of nodal values

Interpolation: Review of historic approaches

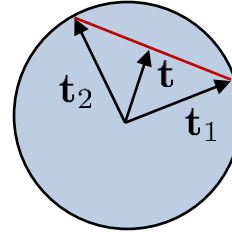
(for non-linear Reissner-Mindlin shell formulations)

Angles

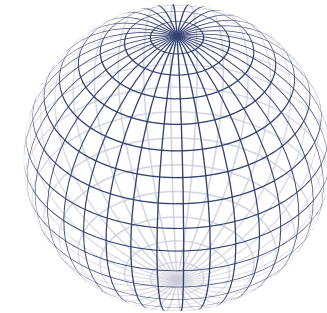
- The unit sphere can be parameterized with an angle pair $(\alpha, \beta) \rightarrow \mathbf{t} = \mathbf{R}(\alpha, \beta) \mathbf{t}_0$, $\mathbf{R}(\alpha, \beta) \in \mathcal{SO}(3)$
RAMM (1976), ARGYRIS(1982), BAŞAR ET AL(1992),
WRIGGERS & GRUTTMANN (1993)

- The resulting interpolation is, e.g. RAMM (1976)

$$\mathbf{t} = \sum_{A=1}^n N^A(\xi) \mathbf{R}_A(\alpha^A, \beta^A) \mathbf{t}_0^A, \quad \mathbf{t}_A = \mathbf{R}_A \mathbf{t}_0^A$$



- Singularities, violates objectivity, unit length constraint violated



WIKIPEDIA (2008) CC BY-SA 3.0

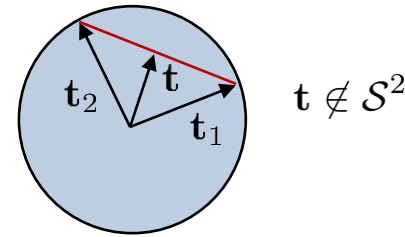
Interpolation: Review of existing approaches

Direct interpolation of the current nodal directors in the embedding space

- Standard interpolation formula for finite elements in vector spaces
 - Simple straightforward interpolation HUGHES & LIU (1981), BATHE & BOLOURCHI (1980), BETSCH & STEINMANN (2002), BENSON ET AL (2010)

$$\mathbf{t} = \sum_{A=1}^n N^A(\xi) \mathbf{t}_A$$

- Interpolated value **does not lie on the unit sphere**
- **Objective**



Generalized Spherical Linear interpolation (SLERP)

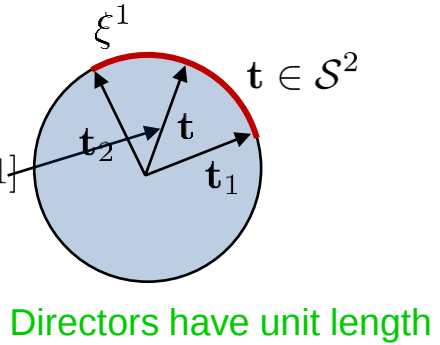
- SLERP: Interpolation between two unit vectors $\mathbf{t}_1, \mathbf{t}_2$

$$\mathbf{t} = \text{SLERP}(\mathbf{t}_1, \mathbf{t}_2, \xi^1) = \frac{\sin((1 - \xi^1) \arccos(\mathbf{t}_1 \cdot \mathbf{t}_2))}{\sin(\arccos(\mathbf{t}_1 \cdot \mathbf{t}_2))} \mathbf{t}_1 + \frac{\sin(\xi^1 \arccos(\mathbf{t}_1 \cdot \mathbf{t}_2))}{\sin(\arccos(\mathbf{t}_1 \cdot \mathbf{t}_2))} \mathbf{t}_2, \quad \xi^1 = [0, 1]$$

- Generalization for four vectors in 2D

$$\mathbf{t} = \text{SLERP}[\text{SLERP}(\mathbf{t}_1, \mathbf{t}_2, \xi^1), \text{SLERP}(\mathbf{t}_3, \mathbf{t}_4, \xi^1), \xi^2] \text{ AREIAS (2013)}$$

- Complicated interpolation
- Objective



Interpolation: Desired Properties

- **Objective**
- **Singularity-free**
- **Useable for arbitrary polynomial order**
- **Invariant to node numbering**
- **Unit length** in the domain

**Suitable
interpolation schemes
for the director field**

Geodesic Finite Elements SANDER (2012)

- GFE define a class of finite elements to interpolate on manifold

- Consider the following interpolation scheme

$$\mathbf{x}^*(\boldsymbol{\xi}, \mathbf{x}_i) = \arg \min_{\mathbf{x} \in \mathbb{R}^n} f(\mathbf{x}, \boldsymbol{\xi}; \mathbf{x}_i) \quad f(\mathbf{x}, \boldsymbol{\xi}; \mathbf{x}_i) = \sum_{i=1}^n N^i(\boldsymbol{\xi}) \|\mathbf{x}_i - \mathbf{x}\|^2$$

Identical to standard interpolation!

$$\bullet \text{ Since: } \frac{\partial f(\mathbf{x}, \boldsymbol{\xi}; \mathbf{x}_i)}{\partial \mathbf{x}} \stackrel{!}{=} \mathbf{0} = \sum_{i=1}^n N^i(\boldsymbol{\xi}) (\mathbf{x}_i - \mathbf{x}) = \sum_{i=1}^n N^i(\boldsymbol{\xi}) \mathbf{x}_i - \underbrace{\sum_{i=1}^n N^i}_{=1} \mathbf{x} \longrightarrow \mathbf{x}^* = \sum_{i=1}^n N^i(\boldsymbol{\xi}) \mathbf{x}_i$$

- Euclidean distance can be generalized for the manifold $\mathcal{M}$

$$\mathbf{x}^* = \arg \min_{\mathbf{x} \in \mathcal{M}} \sum_{i=1}^n N^i(\boldsymbol{\xi}) \text{dist}(\mathbf{x}_i, \mathbf{x})_{\mathcal{M}}^2 = \mathbf{x}_{GP}$$

- objective since distances are a priori rotational invariant
- Directors have unit length
- Implicit interpolation by minimization problem $\rightarrow$ Nonlinear minimization problem at each integration point

Projection-Based Finite Elements GROHS ET AL (2019)

- Projection-based interpolation is a special kind of geodesic finite elements

- In contrast to the GFE definition

$$\mathbf{x}^* = \arg \min_{\mathbf{x} \in \mathcal{M}} \sum_{i=1}^n N^i(\boldsymbol{\xi}) \text{dist}(\mathbf{x}_i, \mathbf{x})_{\mathcal{M}}^2 = \mathbf{x}_{GP}$$

- PB finite elements use the distance of the embedding space

$$\mathbf{x}^* = \arg \min_{\mathbf{x} \in \mathcal{M}} \sum_{i=1}^n N^i(\boldsymbol{\xi}) \text{dist}(\mathbf{x}_i, \mathbf{x})_{\mathbb{R}^n}^2 = \mathbf{x}_{GP}$$

- **objective** since distances are a priori rotational invariant
 - **Directors have unit length**
 - **Implicit interpolation by minimization problem**

What does all that mean for the unit sphere?

Example with two directors:

- **NFE/Standard interpolation:** $N^1(\xi) = 1 - \xi, \quad N^2(\xi) = \xi$

$$\blacksquare \mathbf{t}_{\text{GP}} = \sum_{i=1}^n N^i(\xi) \mathbf{t}_i = \arg \min_{\mathbf{t} \in \mathbb{R}^n} \sum_{i=1}^n N^i(\xi) \|\mathbf{t}_i - \mathbf{t}\|^2$$

- **Projection-Based (PBFE):** GROHS ET AL (2019)

$\mathbf{t}_1$ $\mathbf{t}_2$

$$\begin{aligned} \blacksquare \mathbf{t}_{\text{GP}} &= \arg \min_{\mathbf{t} \in S^2} \sum_{i=1}^n N^i(\xi) \text{dist}(\mathbf{t}_i, \mathbf{t})_{\mathbb{R}^n}^2 \\ &= \arg \min_{\mathbf{t} \in S^2} \sum_{i=1}^n N^i(\xi) \|\mathbf{t}_i - \mathbf{t}\|^2 = \frac{\sum_{i=1}^n N^i(\xi) \mathbf{t}_i}{\|\sum_{i=1}^n N^i(\xi) \mathbf{t}_i\|} \end{aligned}$$

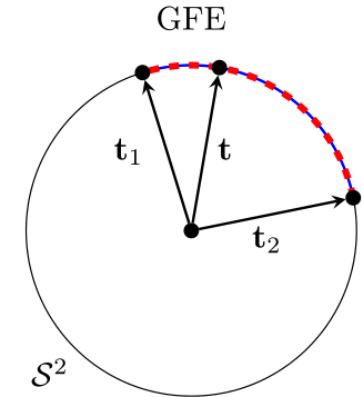
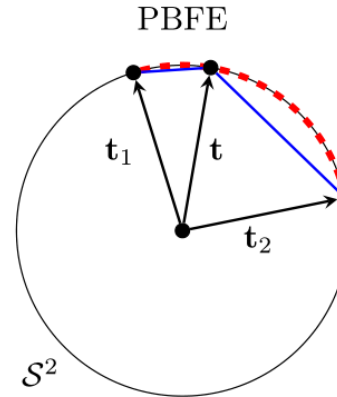
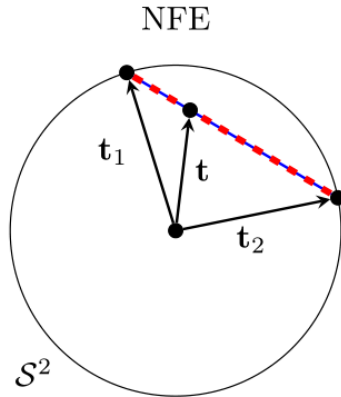
- **Geodesic Finite Elements (GFE):** SANDER (2012)

$$\begin{aligned} \blacksquare \mathbf{t}_{\text{GP}} &= \arg \min_{\mathbf{t} \in S^2} \sum_{i=1}^n N^i(\xi) \text{dist}(\mathbf{t}_i, \mathbf{t})_{S^2}^2 \\ &= \arg \min_{\mathbf{t} \in S^2} \sum_{i=1}^n N^i(\xi) \arccos(\mathbf{t}_i \cdot \mathbf{t})^2 \end{aligned}$$

Suitable interpolation schemes for the director field

What does all that mean for the unit sphere?

Example with two directors:



$$\begin{aligned} \mathbf{t}_{\text{GP}} &= \arg \min_{\mathbf{t} \in \mathbb{R}^n} \sum_{i=1}^n N^i(\boldsymbol{\xi}) \|\mathbf{t}_i - \mathbf{t}\|^2 \\ &= \sum_{i=1}^n N^i(\boldsymbol{\xi}) \mathbf{t}_i \end{aligned}$$

$$\begin{aligned} \mathbf{t}_{\text{GP}} &= \arg \min_{\mathbf{t} \in S^2} \sum_{i=1}^n N^i(\boldsymbol{\xi}) \|\mathbf{t}_i - \mathbf{t}\|^2 \\ &= \frac{\sum_{i=1}^n N^i(\boldsymbol{\xi}) \mathbf{t}_i}{\|\sum_{i=1}^n N^i(\boldsymbol{\xi}) \mathbf{t}_i\|} \end{aligned}$$

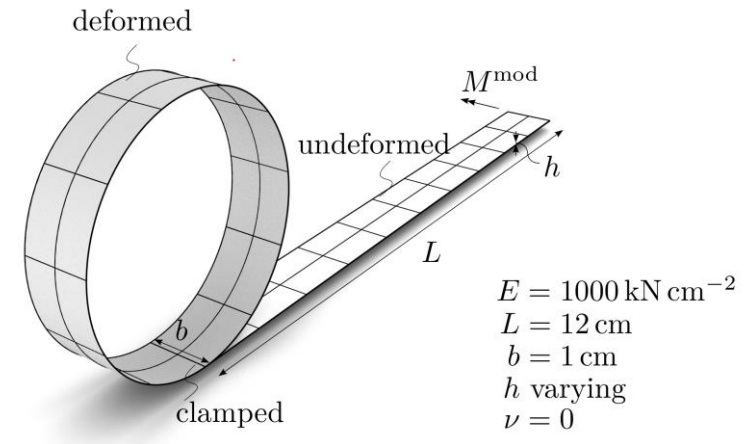
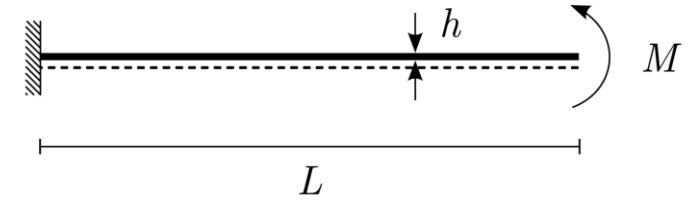
$$\mathbf{t}_{\text{GP}} = \arg \min_{\mathbf{t} \in S^2} \sum_{i=1}^n N^i(\boldsymbol{\xi}) \arccos(\mathbf{t}_i \cdot \mathbf{t})^2$$

Comparison of interpolation schemes

Interpolation: Numerical experiments

Roll-up of clamped beam

- 16 Iterations of Newton's method to reach equilibrium



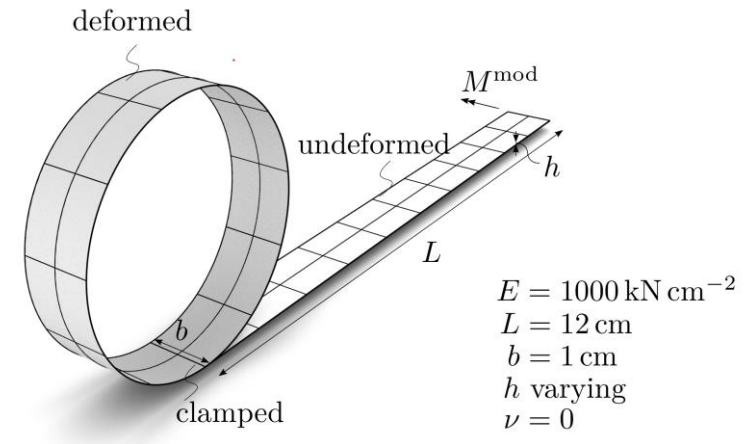
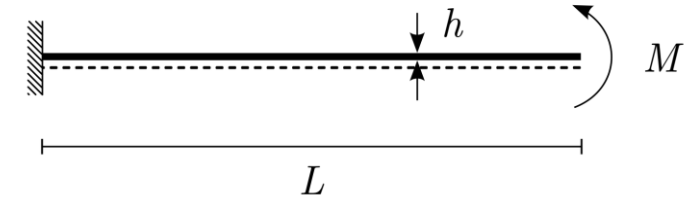
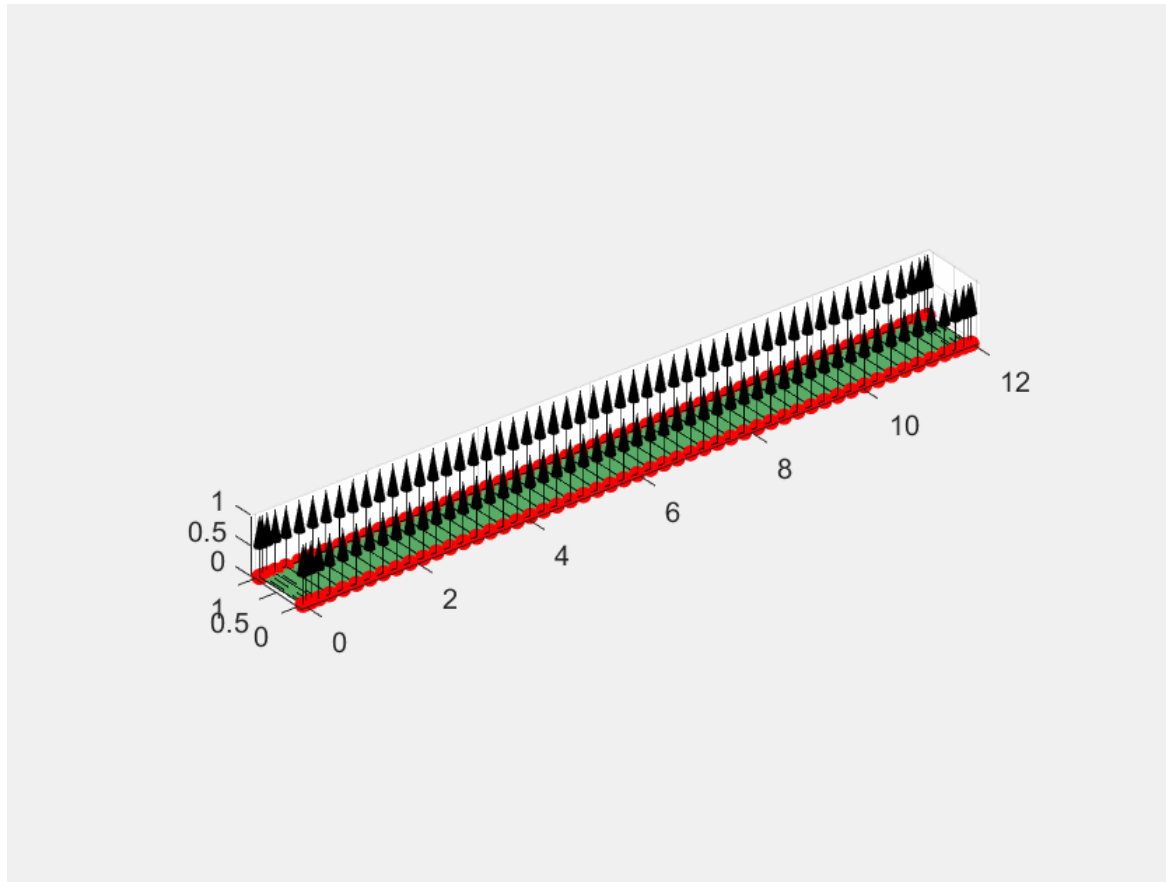
$$\Pi(\varphi, \mathbf{t}) = \int_{\mathcal{B}} \psi_{\text{strain}}(\varphi, \mathbf{t}) \, dV + \Pi_{\text{ext}}(\varphi, \mathbf{t})$$

$$\{\varphi^*, \mathbf{t}^*\} = \arg \left\{ \min_{\varphi \in \mathbb{R}^d} \min_{\mathbf{t} \in \mathcal{S}^{d-1}} \Pi(\varphi, \mathbf{t}) \right\}.$$

Interpolation: Numerical experiments

Roll-up of clamped beam

- 16 Iterations of Newton's method to reach equilibrium



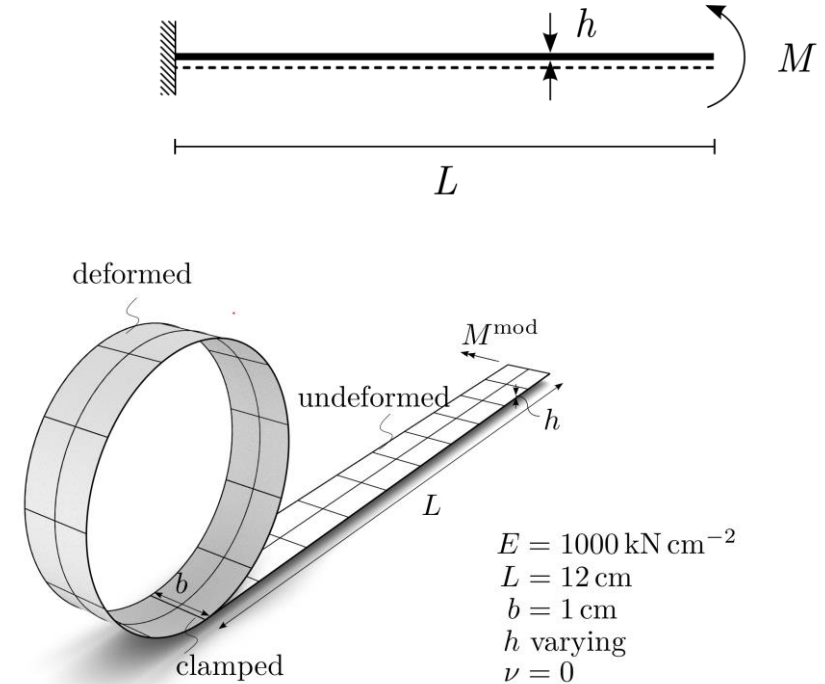
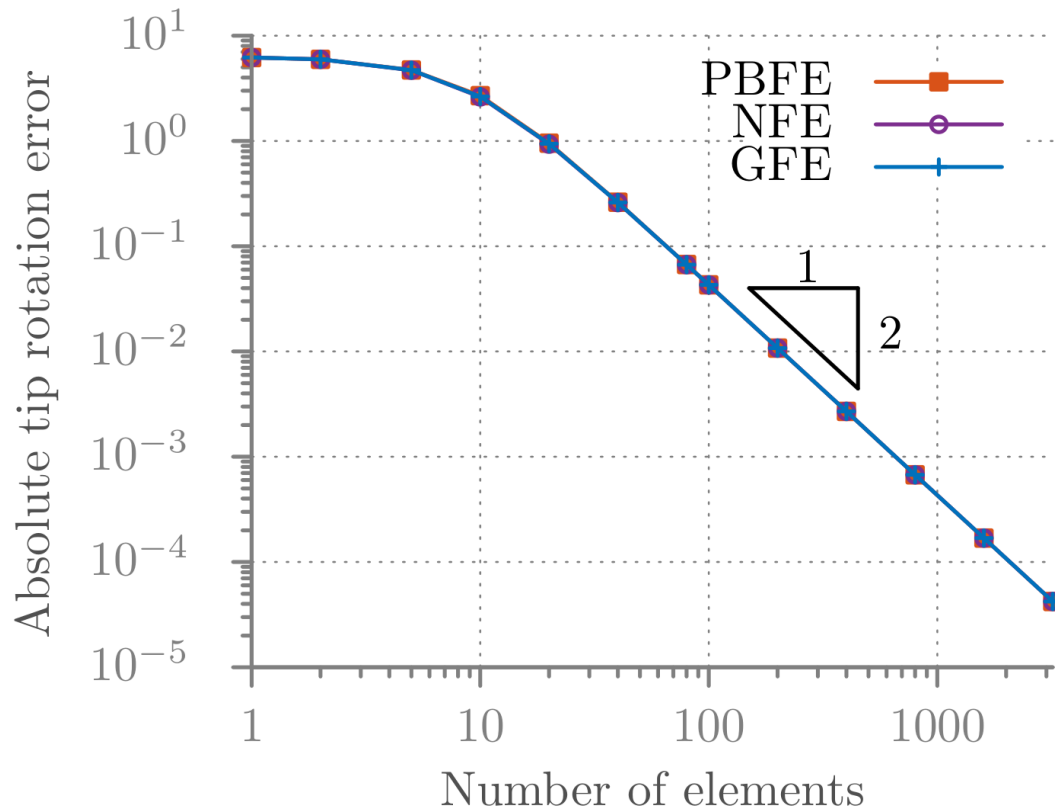
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Interpolation: Numerical experiments

Roll-up of clamped beam

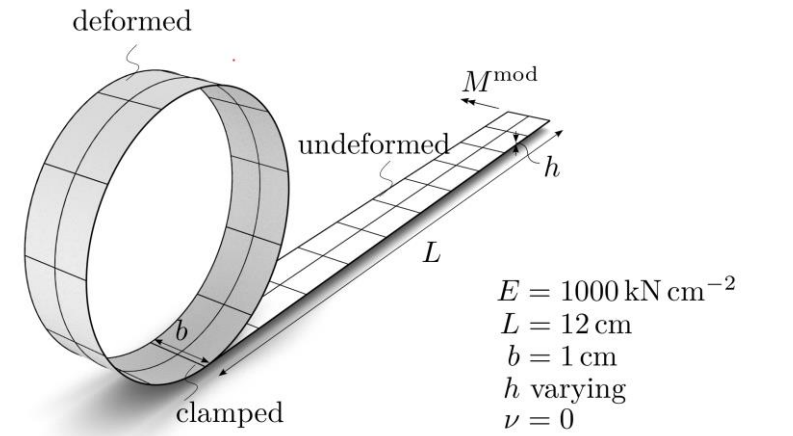
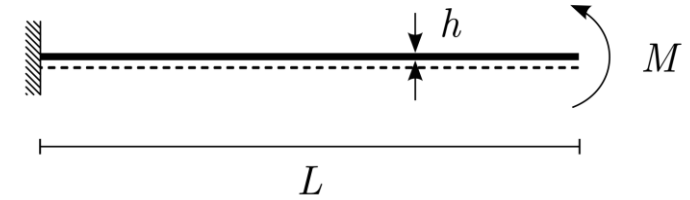
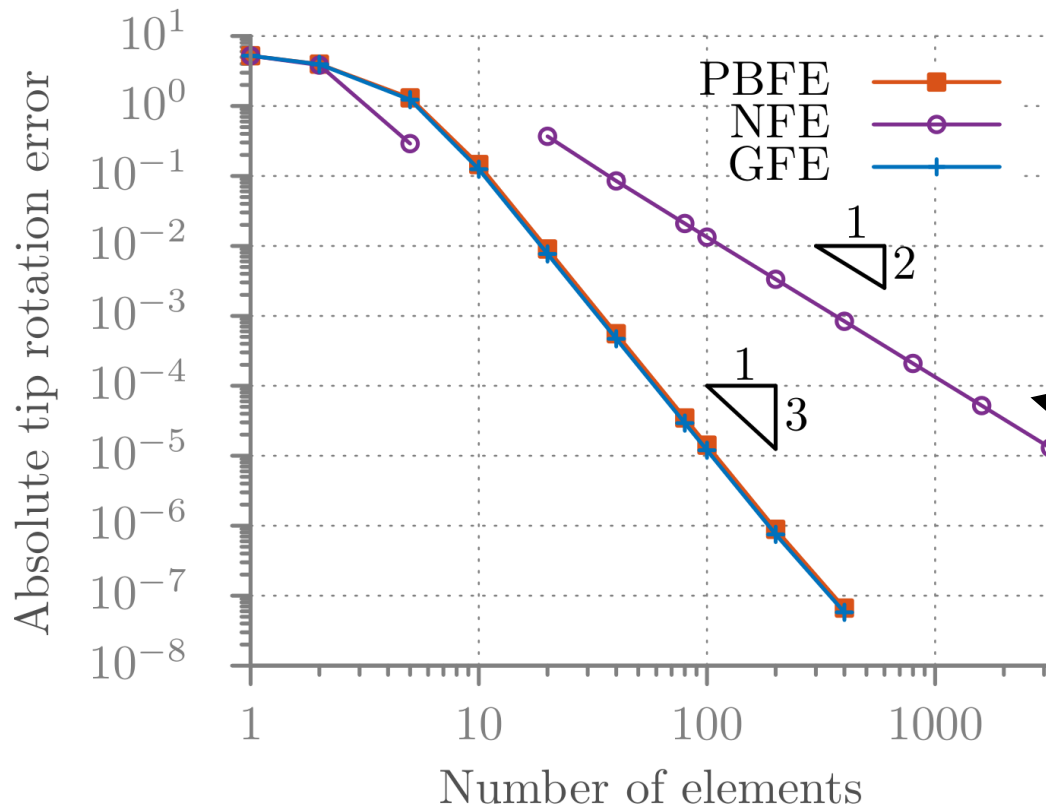
- Reference plane $\rightarrow$ Reference interpolation identical
- Q1 shell elements, C^0 -continuity between elements
- Moderate initial slenderness $L/h = 12 \rightarrow$ Moderate locking



Interpolation: Numerical experiments

Roll-up of clamped beam

- Reference plane $\rightarrow$ Reference interpolation identical
- Quadratic $p = 2$ B-spline shell elements, C^1 -continuity between elements
- Moderate initial slenderness $L/h = 12 \rightarrow$ Moderate locking

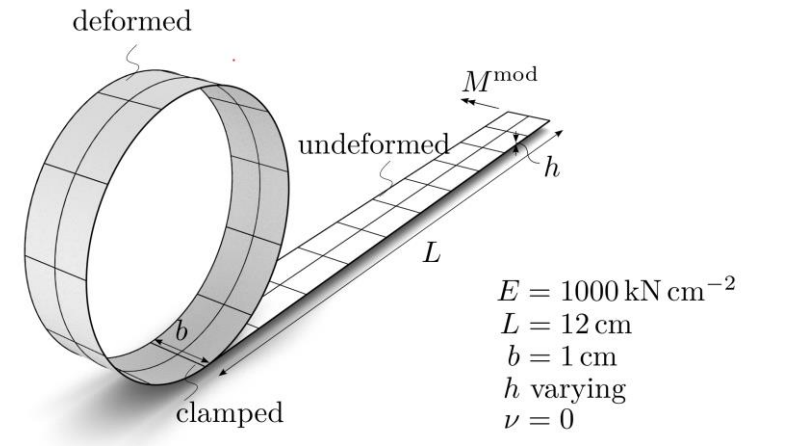
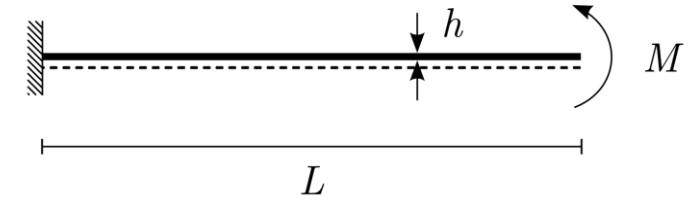
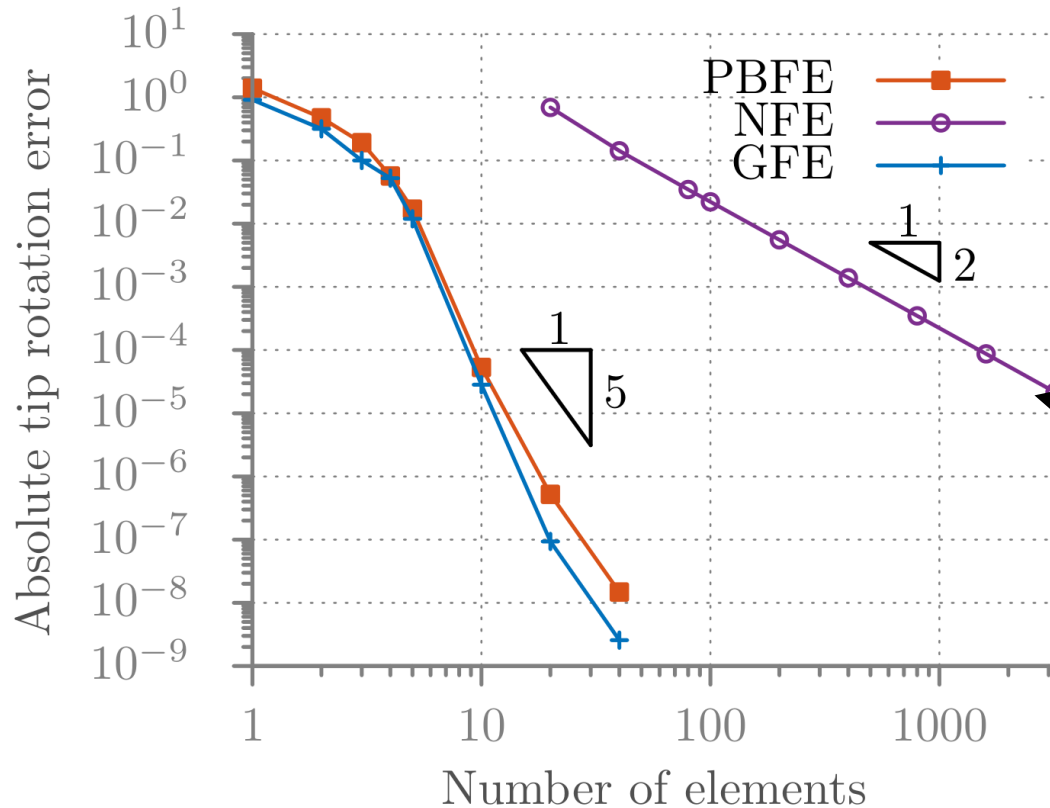


Convergence order degenerates!

Interpolation: Numerical experiments

Roll-up of clamped beam

- Reference plane $\rightarrow$ Reference Interpolation identical
- Quartic $p = 4$ B-spline shell elements, C^3 -continuity between elements
- Moderate initial slenderness $L/h = 12 \rightarrow$ Moderate locking



Convergence order degenerates!

Challenges

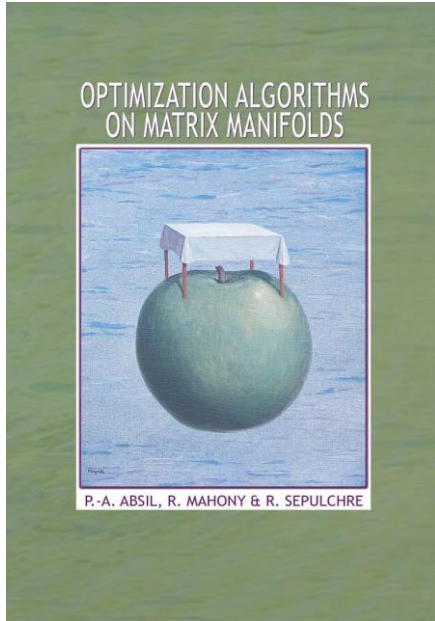
Interpolation on the unit sphere

(Algebraic) optimization on manifolds

Update of nodal values

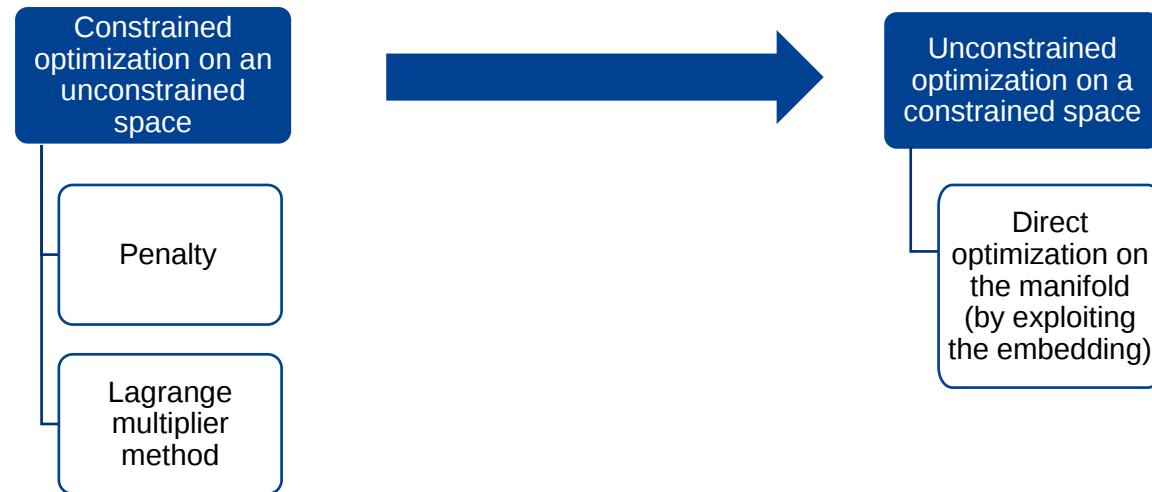
Algebraic optimization on manifolds

Literature

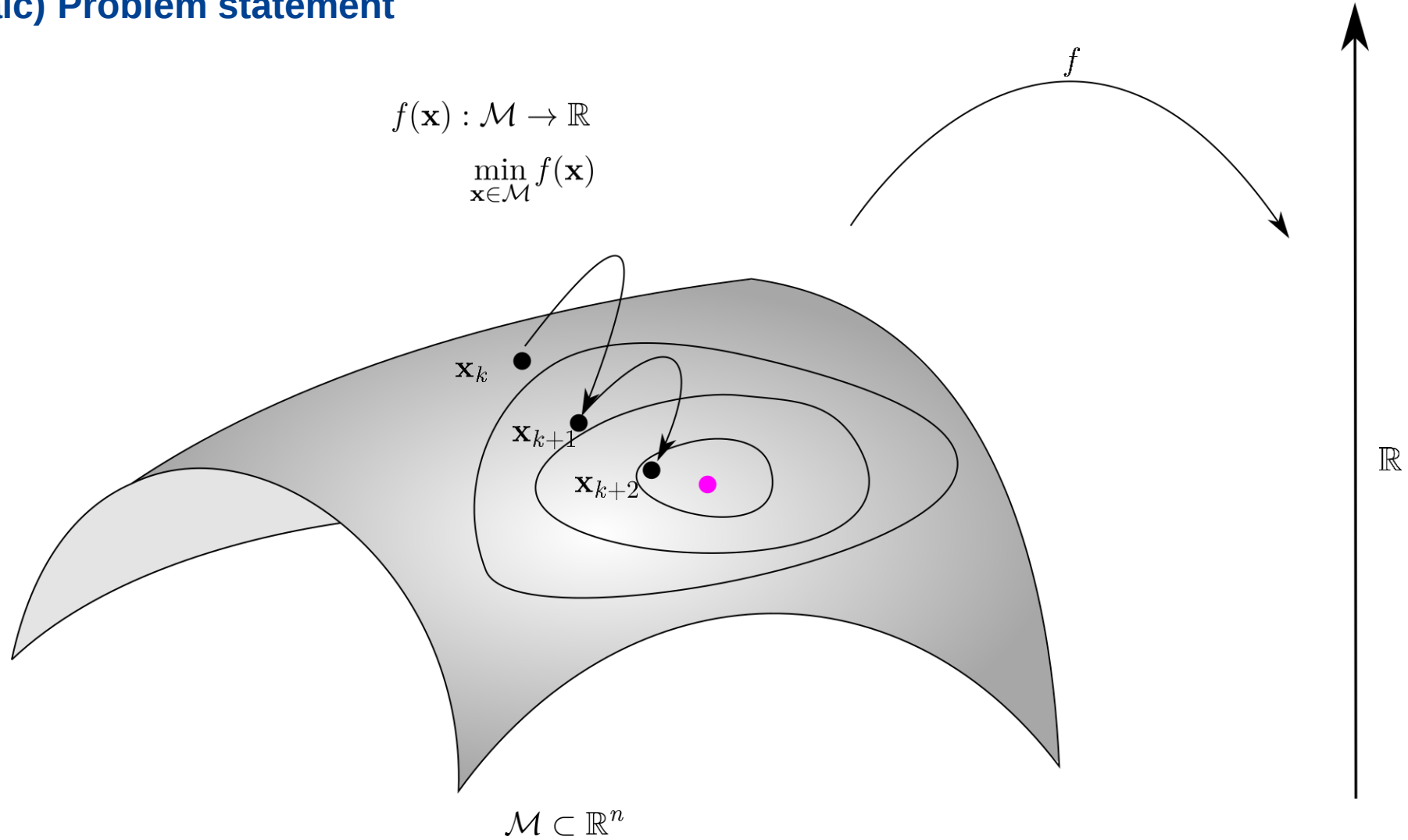


ABSIL PA, MAHONY R, SEPULCHRE R (2008) OPTIMIZATION ALGORITHMS ON MATRIX MANIFOLDS. PRINCETON UNIVERSITY PRESS, DOI: [10.1515/9781400830244](https://doi.org/10.1515/9781400830244)

BOUMAL N (2020) AN INTRODUCTION TO OPTIMIZATION ON SMOOTH MANIFOLDS. AVAILABLE ONLINE, [LINK](#)



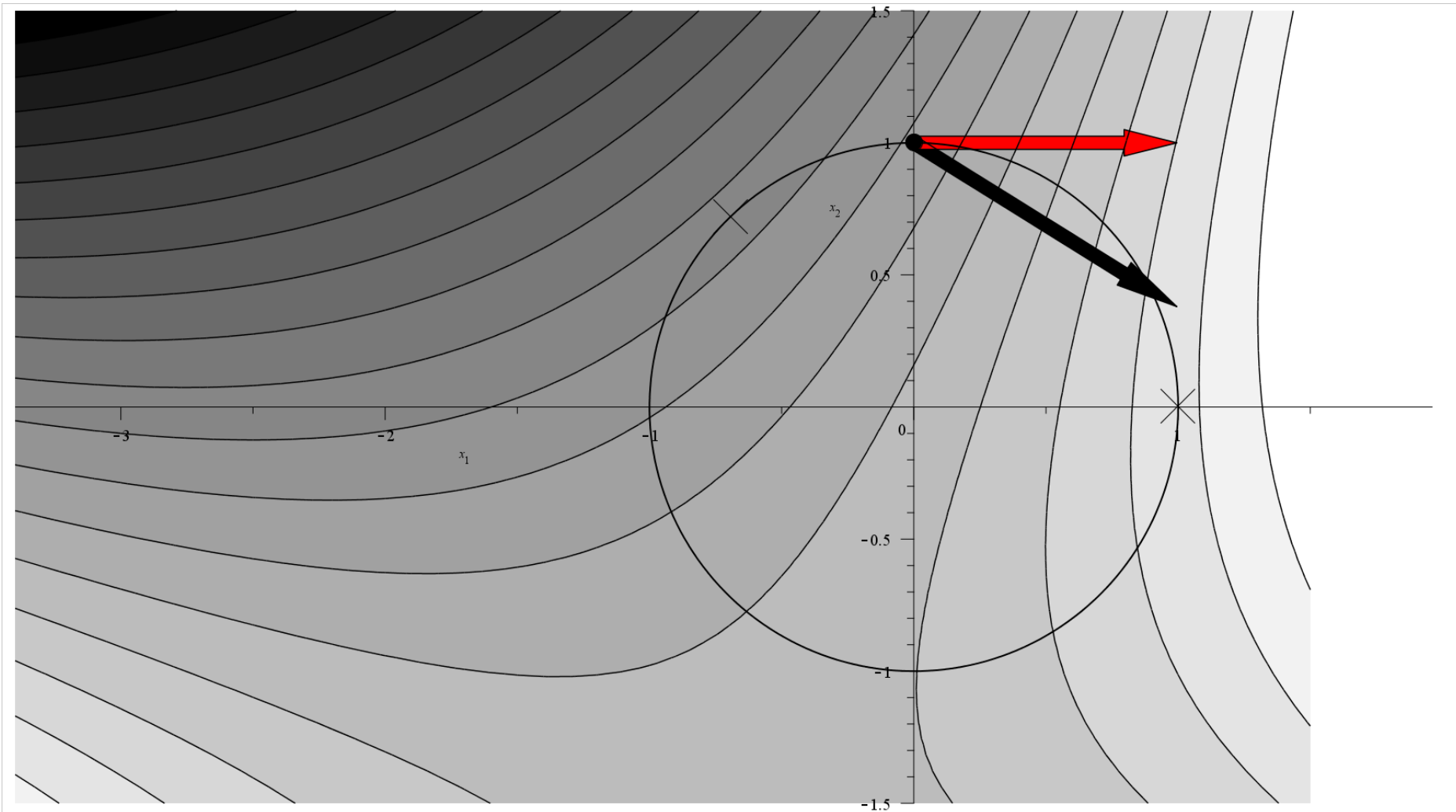
(Algebraic) Problem statement



Riemannian gradient

(exploiting embedding information)

Toy problem, gradient and Riemannian gradient



Riemannian Gradient: submanifolds

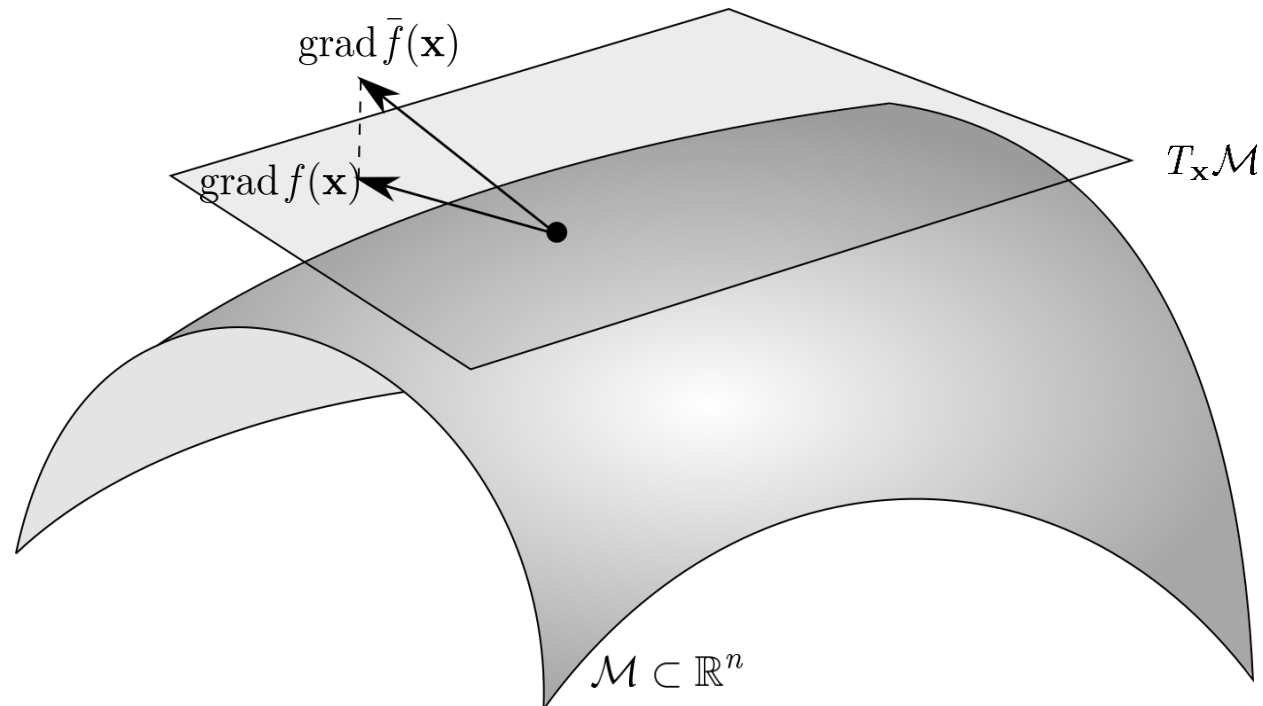
$$f(\mathbf{x}) : \mathcal{M} \rightarrow \mathbb{R}$$

$$\bar{f}(\mathbf{x}) : \mathbb{R}^n \rightarrow \mathbb{R}$$

“For Riemannian submanifolds, the Riemannian gradient is the orthogonal projection of the “classical” gradient to the tangent spaces.”

BOUMAL N (2020) AN INTRODUCTION TO OPTIMIZATION ON SMOOTH MANIFOLDS.

$$\text{grad } f(\mathbf{x}) = P_{\mathbf{x}} \text{grad } \bar{f}(\mathbf{x})$$



- No charts
- No artificial singularities
- Simple linearization

Challenges

Interpolation on the unit sphere

(Algebraic) optimization on manifolds

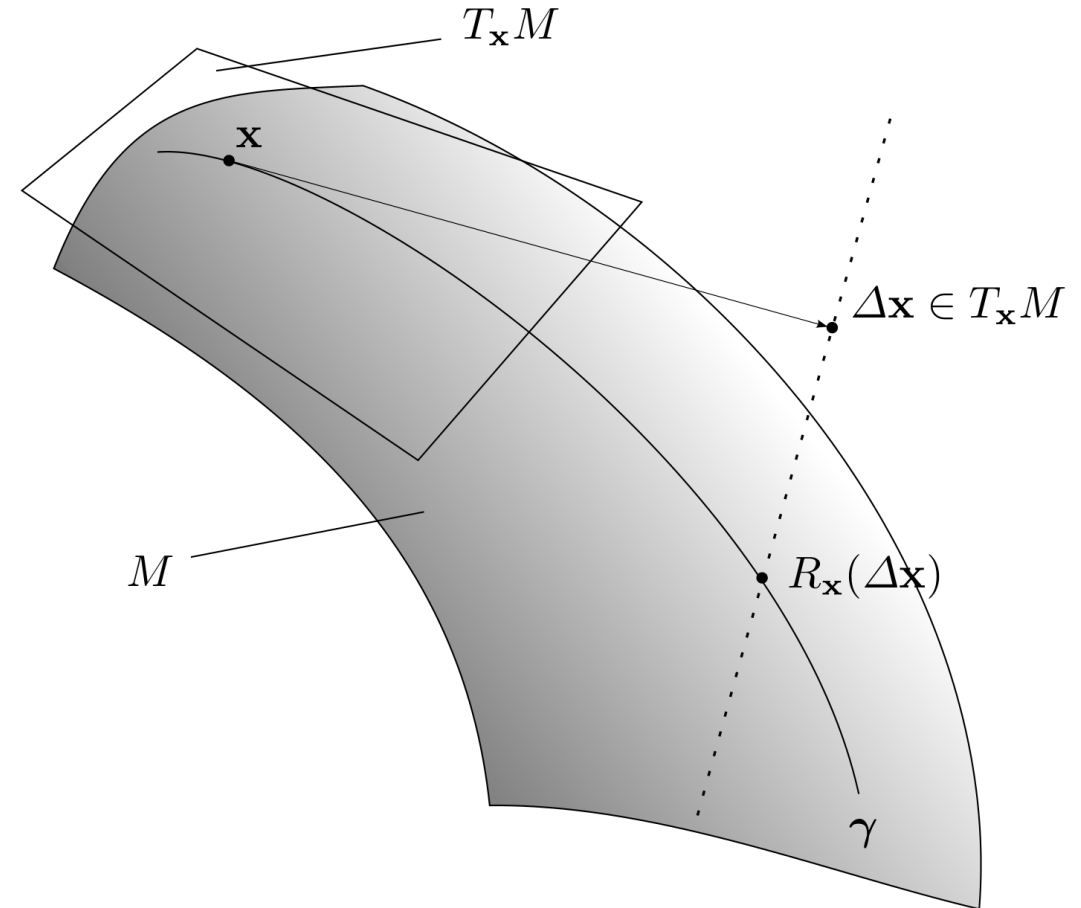
Update of nodal values

**Update of
nodal values**

Update of nodal values

$$\mathbf{x}_k + \Delta \mathbf{x}_k \notin \mathcal{M}$$

Geodesics generalize the concept of straight lines



The exponential map creates the *unique* geodesic curve starting at $\mathbf{x}_k$ in direction $\Delta \mathbf{x}_k$ with constant speed

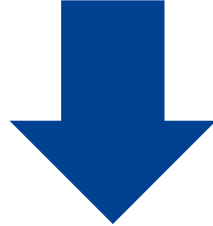
$$\gamma(t) = \exp_{\mathbf{x}_k}(t\Delta \mathbf{x}_k) \quad \longrightarrow \quad \mathbf{x}_{k+1} = \exp_{\mathbf{x}_k}(\Delta \mathbf{x}_k)$$

Along these geodesics one could perform e.g. line search

Update of nodal values

ABSIL PA, "OPTIMIZATION ON MANIFOLDS: METHODS AND APPLICATIONS", LEUVEN, 18 SEP 2009.

Luenberger (1973), Introduction to linear and nonlinear programming.
Luenberger mentions the idea of performing line search along geodesics, "*which we would use if it were computationally feasible (which it definitely is not)*".



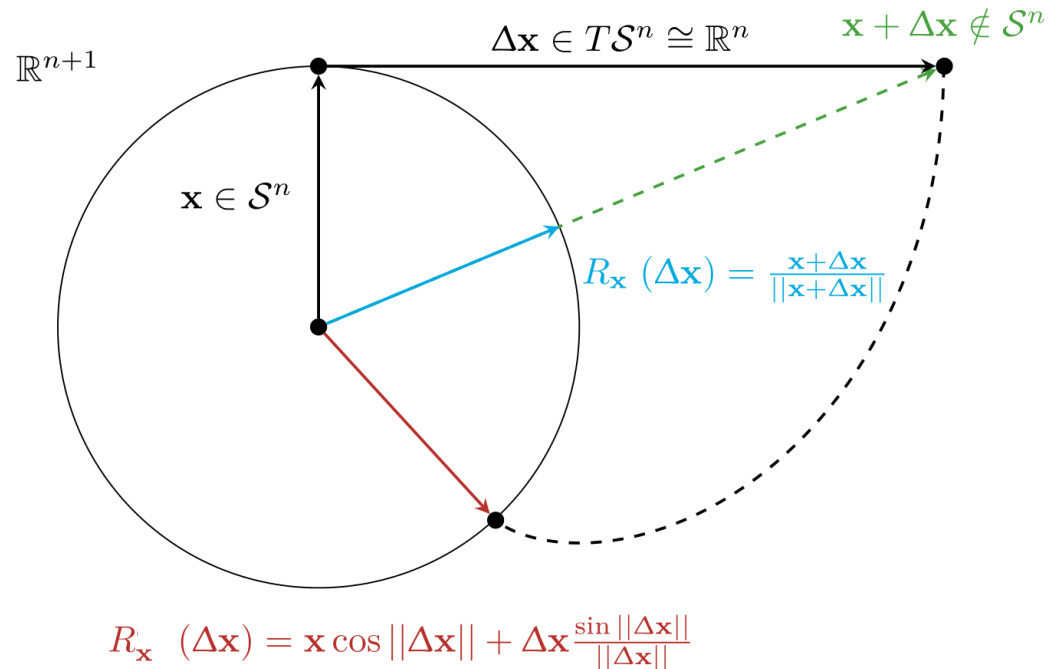
Generalize the concept of the exponential map $\rightarrow$ Retractions

Update of nodal values

Retractions for the unit sphere

$$R_{\mathbf{x}}^{\text{exp}}(\Delta \mathbf{x}) = \exp_{\mathbf{x}}(\Delta \mathbf{x}) = \cos(\|\Delta \mathbf{x}\|)\mathbf{x} + \frac{\sin(\|\Delta \mathbf{x}\|)}{\|\Delta \mathbf{x}\|}\Delta \mathbf{x}$$

$$R_{\mathbf{x}}^{\text{rrn}}(\Delta \mathbf{x}) = \frac{\mathbf{x} + \Delta \mathbf{x}}{\|\mathbf{x} + \Delta \mathbf{x}\|}$$

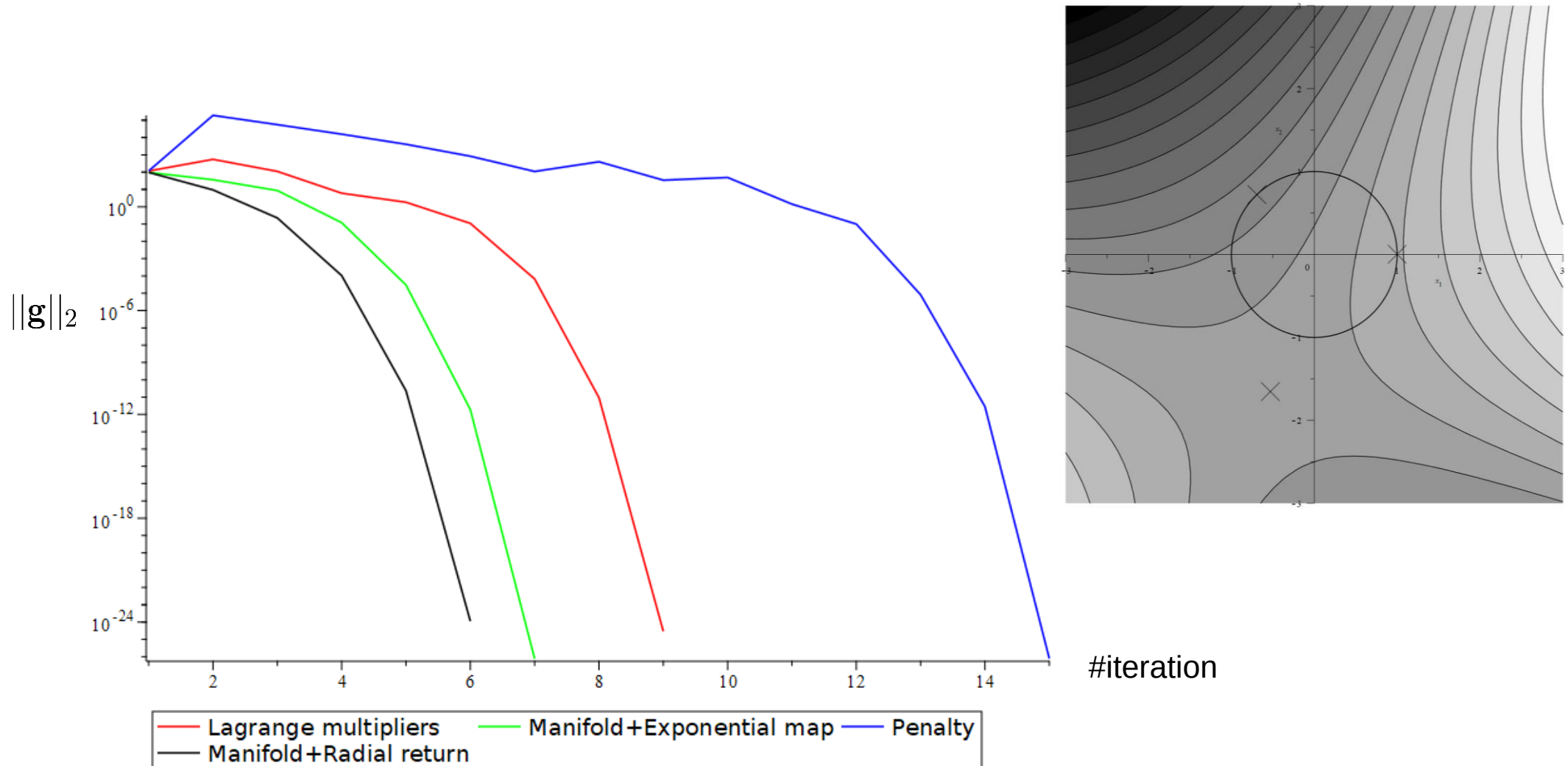


Algebraic optimization on manifolds

Update of nodal values

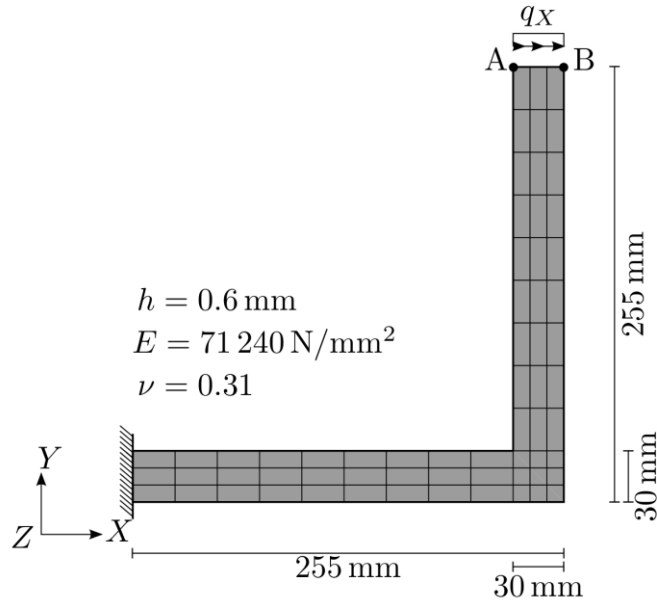
Retractions for the unit sphere

Toy problem: Newton's method, iteration count vs. gradient norm



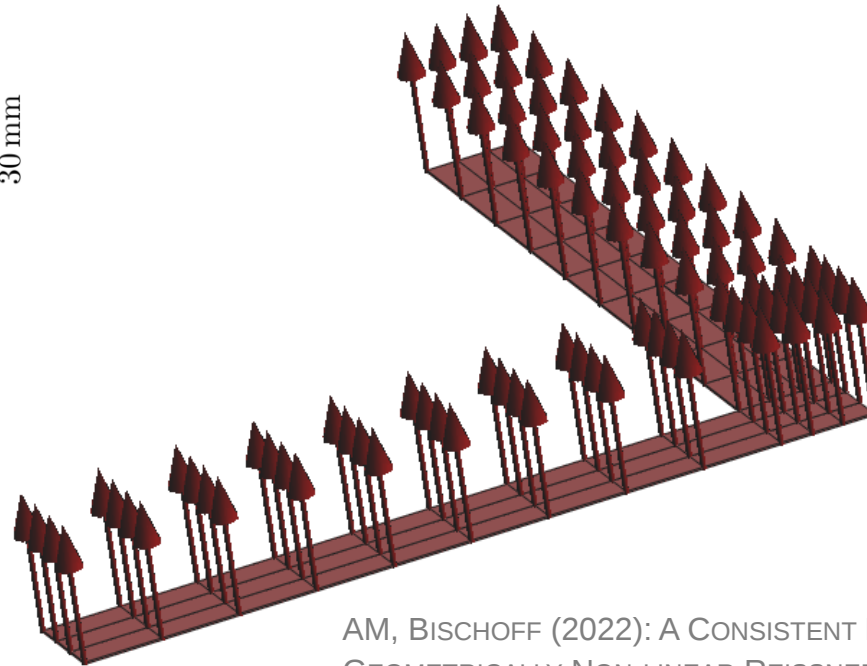
Simulations

Simulation elastic deformation of shells



$$\Pi(\varphi, \mathbf{t}) = \int_{\mathcal{B}} \psi_{\text{strain}}(\varphi, \mathbf{t}) \, dV + \Pi_{\text{ext}}(\varphi, \mathbf{t})$$

$$\{\varphi^*, \mathbf{t}^*\} = \arg \left\{ \min_{\varphi \in \mathbb{R}^d} \min_{\mathbf{t} \in \mathcal{S}^{d-1}} \Pi(\varphi, \mathbf{t}) \right\}.$$

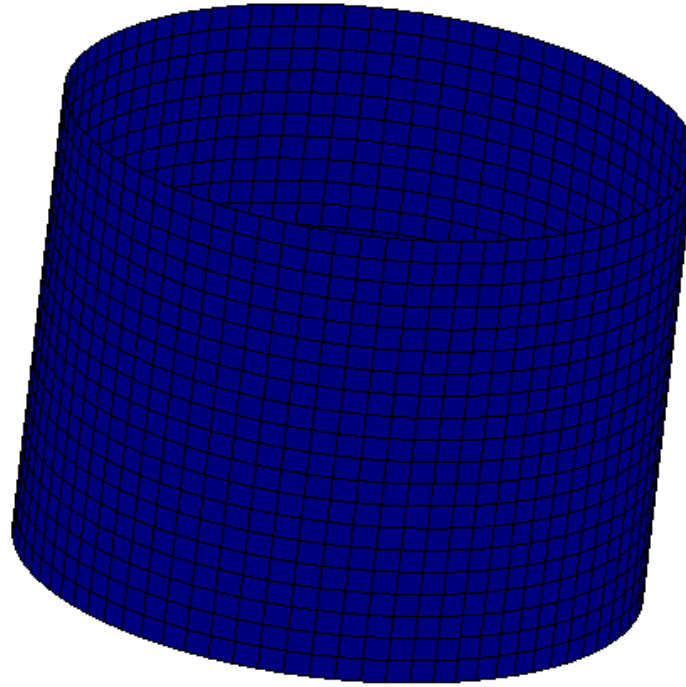


AM, BISCHOFF (2022): A CONSISTENT FINITE ELEMENT FORMULATION OF THE GEOMETRICALLY NON-LINEAR REISSNER-MINDLIN SHELL MODEL, [DOI](#)

Simulation of magnetic vortices

Simulation elastic deformation of shells

Riemannian Trust-Region method



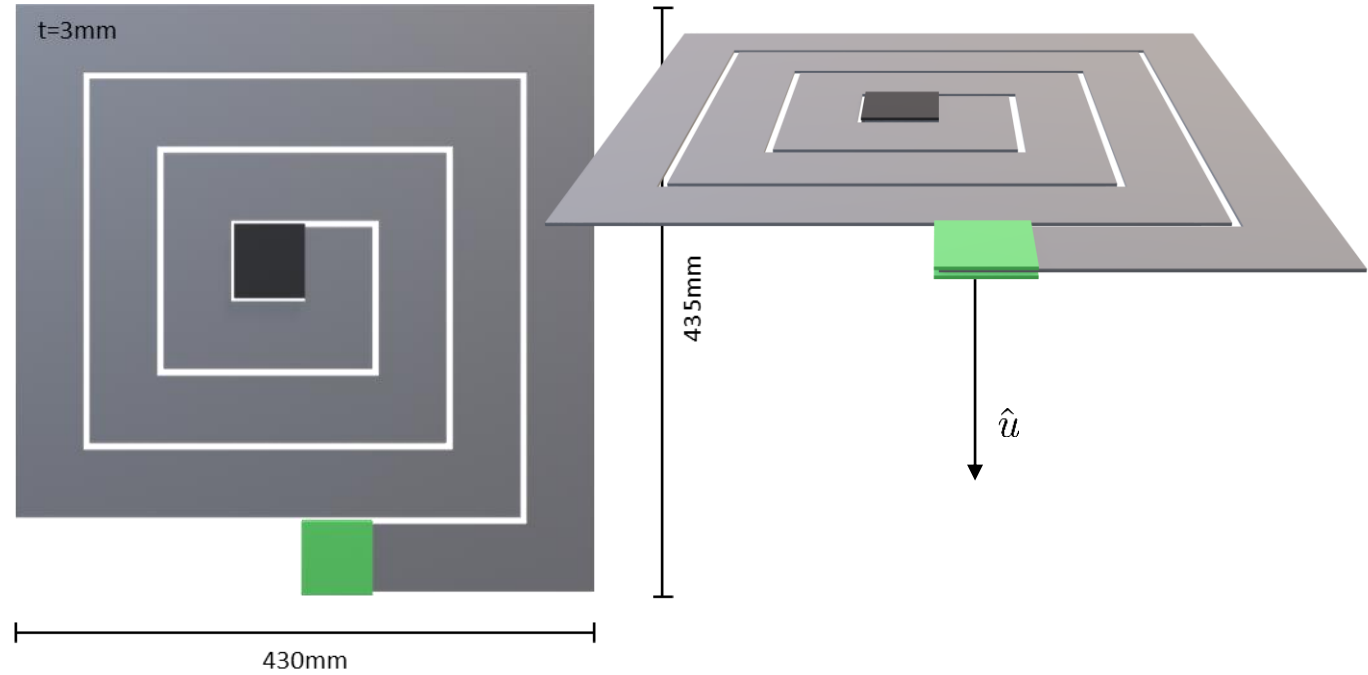
Minimizers for cylinder buckling

Simulation of Reissner-Mindlin shells

Simulation elastic deformation of shells

$$E = 210000$$

$$\nu = 0.3$$



Simulation of Reissner-Mindlin shells

Simulation elastic deformation of shells

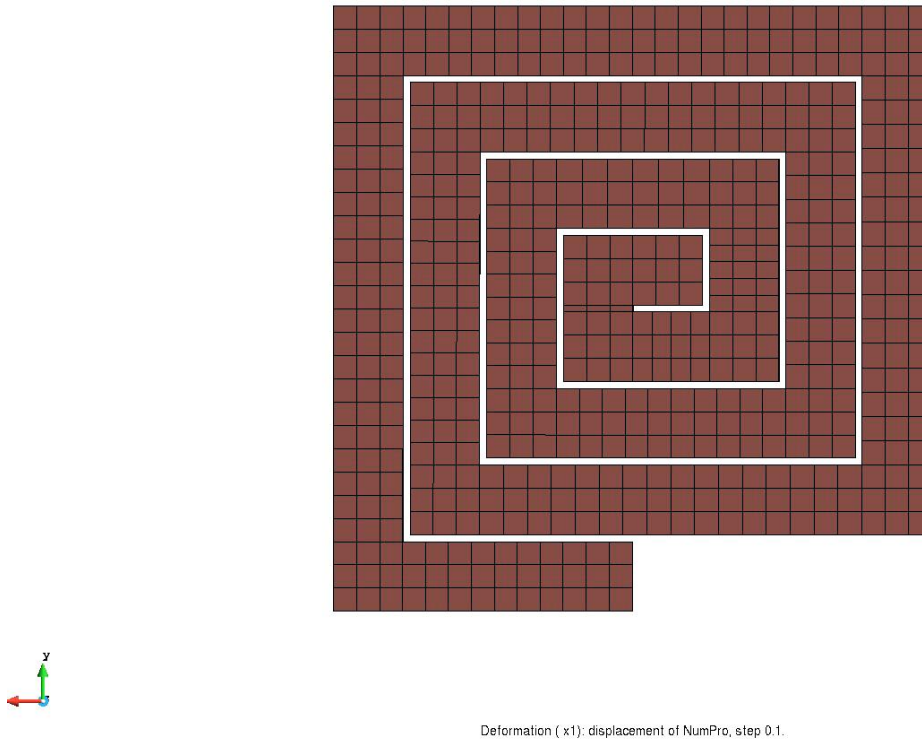
50 load steps



Simulation of Reissner-Mindlin shells

Simulation elastic deformation of shells

Several load steps



Simulation of micromagnetics

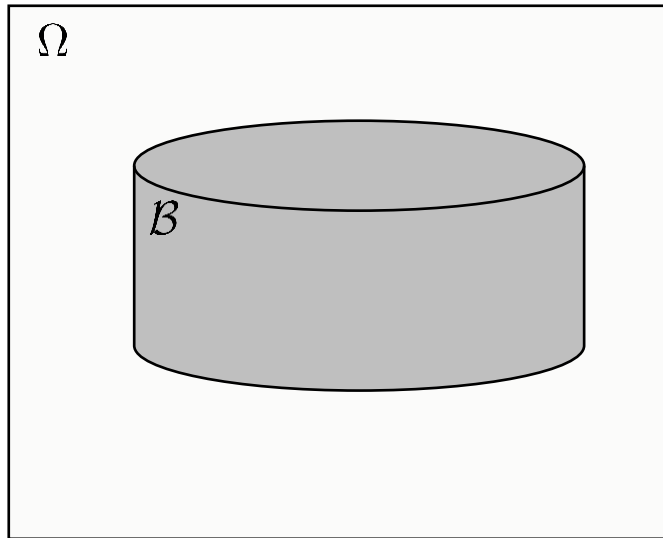
Maxwell's equation in vacuum and matter

$$\begin{aligned}\Omega \quad & \mathbf{B} = \mu_0 \mathbf{H} \\ & \operatorname{div} \mathbf{B} = 0 \\ & \nabla \times \mathbf{H} = 0\end{aligned}$$

$$\begin{aligned}\mathcal{B} \quad & \mathbf{B} = \mu_0 (\mathbf{M} + \mathbf{H}) \\ & \operatorname{div} \mathbf{B} = 0 \\ & \nabla \times \mathbf{H} = 0\end{aligned}$$

Simulation of micromagnetics

$$\Pi(\mathbf{a}, \mathbf{m}) = \int_{\Omega \setminus \mathcal{B}} \frac{1}{2} \|\operatorname{curl} \mathbf{a}\|^2 dV + \int_{\mathcal{B}} \frac{1}{2} \|\operatorname{grad} \mathbf{m}\|^2 + \frac{1}{2} \|\operatorname{curl} \mathbf{a}\|^2 - \operatorname{curl} \mathbf{a} \cdot \mathbf{m} dV$$

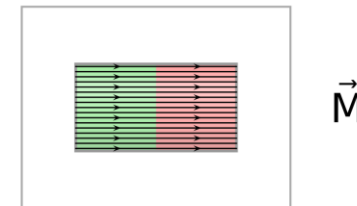
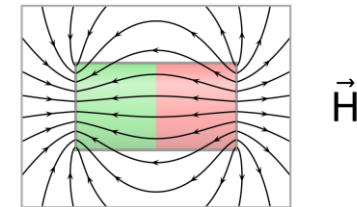
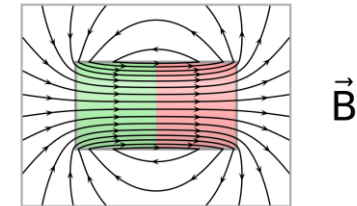


$$\mathbf{m} \in \mathcal{S}^2$$

$$\mathbf{a} \in \mathbb{R}^3$$

$$\mathbf{b} = \nabla \times \mathbf{a}$$

$$\min_{\mathbf{m} \in \mathcal{S}^2} \min_{\mathbf{a} \in \mathbb{R}^3} \Pi(\mathbf{m}, \mathbf{a})$$



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Simulation of micromagnetics



Summary

- Historical approaches can be outperformed by interpreting the problem as optimization on manifolds
- Interpolation must stay on the manifolds
- Lots of customization points exist, e.g., retractions
- Many other manifolds can be found in literature
- Methods can be used for other physical simulations (micromagnetics)

Not mentioned:

Examples:

The important Sobolev space $W^{1,2}(\Omega, \mathcal{M})$ does not even always possess the structure of a Banach manifold
Geodesics for interpolation are not always unique

Algebraic consideration of optimization of manifolds:

Rosen JB (1961) *The gradient projection method for nonlinear programming. Part II. nonlinear constraints*. SIAM 9(4):514–532, doi:[10.1137/0109044](https://doi.org/10.1137/0109044)

Luenberger, David G. (1973) *Introduction to linear and nonlinear programming*. Vol. 28. Reading, MA: Addison-wesley,.

Adler RL, Dedieu JP, Margulies JY, Martens M, Shub M (2002) *Newton's method on Riemannian manifolds and a geometric model for the human spine*. IMA J Numer Anal 22(3):359–390, doi:[10.1093/imanum/22.3.359](https://doi.org/10.1093/imanum/22.3.359)

Absil PA, Mahony R, Sepulchre R (2008) *Optimization Algorithms on Matrix Manifolds*. Princeton University Press, doi:[10.1515/9781400830244](https://doi.org/10.1515/9781400830244)

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Absil PA, Mahony R, Trumpf J (2013) *An extrinsic look at the riemannian hessian*. In: Nielsen F, Barbaresco F (eds) *Geometric Science of Information*, pp 361–368, doi:[10.1007/978-3-642-40020-9_39](https://doi.org/10.1007/978-3-642-40020-9_39)

Huang W (2017)). *Introduction to riemannian bfgs methods*. Available online, URL [Link to online resource](#)

Boumal N (2020) *An introduction to optimization on smooth manifolds*. Available online, URL [Link to online resource](#)

Finite elements for manifolds:

Grohs P (2011) *Finite elements of arbitrary order and quasiinterpolation for data in Riemannian manifolds*. Tech. Rep. 2011-56, Seminar for Applied Mathematics, ETH Zürich, URL https://www.sam.math.ethz.ch/sam_reports/reports_final/reports2011/2011-56.pdf

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Grohs P, Hardering H, Sander O (2015) *Optimal A Priori Discretization Error Bounds for Geodesic Finite Elements*. Found Comput Math 15(6):1357–1411, doi:[10.1007/s10208-014-9230-z](https://doi.org/10.1007/s10208-014-9230-z)

Grohs P, Hardering H, Sander O, Sprecher M (2019) *Projection-based finite elements for nonlinear function spaces*. SIAM J Numer Anal, doi: [10.1137/18M1176798](https://doi.org/10.1137/18M1176798)

Hardering H (2018) *L2-discretization error bounds for maps into Riemannian manifolds*. Numer Math (Heidelb) 139(2):381–410, doi:[10.1007/s00211-017-0941-3](https://doi.org/10.1007/s00211-017-0941-3)

Physical simulations (Small sample):

Sander O, Neff P, Birsan M (2016) *Numerical treatment of a geometrically nonlinear planar Cosserat shell model*. Comput Mech 57(5):817–841, doi:[10.1007/s00466-016-1263-5](https://doi.org/10.1007/s00466-016-1263-5)

Bischoff M, (1999) *Theorie und Numerik einer dreidimensionalen Schalenformulierung*, doi: [10.18419/opus-126](https://doi.org/10.18419/opus-126)

AM, Bischoff M.(2022) *A Consistent Finite Element Formulation of the Geometrically Non-linear Reissner-Mindlin Shell Model*. DOI :[10.1007/s11831-021-09702-7](https://doi.org/10.1007/s11831-021-09702-7).



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Thank you!



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