

An objective and path-independent geometrically non-linear Reissner-Mindlin shell formulation

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31.7.22 - 5.8.22



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Kinematics of a nonlinear Reissner-Mindlin shell model

Kinematic assumptions

Geometry approximation:

 $\mathbf{x} = \mathbf{\Phi}_t(\xi^1,\!\xi^2,\!\xi^3) = \boldsymbol{\varphi}(\xi^1,\!\xi^2) + \xi^3 \mathbf{t}(\xi^1,\!\xi^2)$

- First order transverse shear effects are taken into account:
 - The director field $t(\xi^1,\xi^2)$ 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. $\mathbf{t}: \mathcal{A} \to \mathcal{S}^2 \subset \mathbb{R}^3$ $\mathcal{S}^2 = \{ \mathbf{x} \in \mathbb{R}^3 \mid | \mathbf{x} \cdot \mathbf{x} = 1 \}$

Discretization

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





Director fields



Example 1: Geometrically nonlinear Reissner-Mindlin shell



 $\mathbf{t}(\xi^1,\xi^2)$ $\mathbf{t}:\mathcal{A} o \mathcal{S}^2\subset \mathbb{R}^3$

Director fields



Example 2: Geometrically nonlinear beam





Director fields



Example 3: Magnetic Maxwell equations in matter



$$\mathbf{m}(\xi^1,\xi^2,\xi^3)$$
 $\mathbf{m}:\mathcal{A} o\mathcal{S}^2\subset\mathbb{R}^3$

Outline





Open questions/Outlook



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Interpolation: Review of historic approaches

(for Reissner-Mindlin shell)

Baustatik und Baudynamik

Example 1: Angles

- The unit sphere can be parameterized with an angle pair $(\alpha,\beta) \rightarrow \beta$
 - 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

 $\mathbf{t} = \mathbf{R}(\alpha, \beta) \mathbf{t}_0, \ \mathbf{R}(\alpha, \beta) \in \mathcal{SO}(3)$

RAMM (1976), ARGYRIS(1982), BAŞAR ET AL(1992), WRIGGERS & GRUTTMANN (1993)



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Example 2: Direct interpolation of the nodal directors

- 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





Example 3: Generalized Spherical Linear interpolation (SLERP)

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

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

• Generalization for four vectors in 2D

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

- Complicated interpolation
- Objective



Directors have unit length



Example 4: Interpolating increments

• Only incremental rotations are interpolated SIMO ET AL (1990), BÜCHTER & RAMM (1992), DORNISCH ET AL (2016)

$$\Delta \boldsymbol{\theta} = \sum_{i} N^{i}(\xi, \eta) \Delta \boldsymbol{\theta}_{i}$$

• Rotation matrices are updated at each IP SIMO ET AL (1990), DORNISCH ET AL (2016)

 $\mathbf{R}_{IP}^{k+1} = \exp(\Delta \hat{\boldsymbol{\theta}}) \mathbf{R}_{IP}^{k}$

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→ Non-objective and path-dependent for elastic problems as proven in CRISFIELD & JELENIĆ (1999)

Directors at GPs have unit length

Interpolation: Desired Properties



- Objective
- Singularity-free
- Interpolated value stays on the manifold
- Useable for arbitrary polynomial order
- Generalizable from 1D
- Invariant to node numbering
- No artificial path-dependence



Geodesic Finite Elements SANDER (2012)

GFE define a class of finite elements to interpolate on manifold

Consider the following interpolation scheme

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

$$\text{Identical to standard interpolation!}$$

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

$$\mathbf{x}^{*} = \sum_{i=1}^{n} N^{i}(\boldsymbol{\xi})\mathbf{x}_{i}$$

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

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

- 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

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

• PB finite elements use the distance of the embedding space

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

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

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 \mathbf{t}_2

What does all that mean for the unit sphere?

Example with two directors:

• **NFE**/Standard interpolation:

$$N^{1}(\xi) = 1 - \xi, \quad N^{2}(\xi) = \xi$$

 \mathbf{t}_1

$$\blacksquare \mathbf{t}_{\mathrm{GP}} = \sum_{i=1}^{n} N^{i}(\xi) \mathbf{t}_{i}$$

• Projection-Based (PBFE): Grohs et al (2019) • $\mathbf{t}_{GP} = \arg\min_{\mathbf{t}\in\mathcal{S}^2}\sum_{i=1}^n N^i(\xi)\operatorname{dist}(\mathbf{t}_i,\mathbf{t})^2_{\mathbb{R}^n}$ = $\arg\min_{\mathbf{t}\in\mathcal{S}^2}\sum_{i=1}^n N^i(\xi)||\mathbf{t}_i - \mathbf{t}||^2 = \frac{\sum_{i=1}^n N^i(\xi)\mathbf{t}_i}{||\sum_{i=1}^n N^i(\xi)\mathbf{t}_i||}$

• Geodesic Finite Elements (GFE): Sander (2012) • $\mathbf{t}_{\mathrm{GP}} = \arg\min_{\mathbf{t}\in\mathcal{S}^2}\sum_{i=1}^n N^i(\xi)\operatorname{dist}(\mathbf{t}_i,\mathbf{t})_{\mathcal{S}^2}^2$ $= \arg\min_{\mathbf{t}\in\mathcal{S}^2}\sum_{i=1}^n N^i(\xi)\operatorname{arccos}(\mathbf{t}_i\cdot\mathbf{t})^2$

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What does all that mean for the unit sphere?

Example with two directors:



Comparison of interpolation schemes

Baustatik und Baudynamik

Roll-up of clamped beam



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

Roll-up of clamped beam

- Reference plane → Reference interpolation identical
- Q1 shell elements, C^{0} -continuity between elements





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Roll-up of clamped beam

- Reference plane → Reference interpolation identical
- Quadratic p = 2 B-spline shell elements, C^1 -continuity between elements





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Roll-up of clamped beam

- Reference plane → Reference Interpolation identical
- Quartic p = 4 B-spline shell elements, C^3 -continuity between elements





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Interpolation: Desired Properties



Geometric (PBFE and GFE) satisfy all the desired properties

- Geometric (PBFE and GFE) satisfy all the desired properties
 - Objective
 - Singularity-free
 - Interpolated value stays on the manifold
 - Useable for arbitrary polynomial order
 - Generalizable from 1D
 - Invariant to node numbering
 - No artificial path-dependence
- Drawbacks
 - Maybe expensive to evaluate

Interlude: Function spaces

Interlude: Test function spaces

- First order geometric finite element functions satisfy $V_h^M(\Omega) \subset H^1(\Omega, M)$
- Discrete test functions live in the tangent bundle $\delta \mathbf{u} \in T_{\mathbf{u}} V_h^M(\Omega)$
- Manifold

u
$$V_h^M(\Omega) \subset H^1(\Omega, M)$$

 $\delta \mathbf{u}$ $T_{\mathbf{u}}V_h^M(\Omega)$

• Vector space

$$\mathbf{u} = \sum_{i=1}^{n} N^{i}(\xi) \mathbf{u}_{i} \qquad V_{h}^{\mathbb{R}^{n}}(\Omega) \subset H^{1}(\Omega, \mathbb{R}^{n})$$

$$\delta \mathbf{u} = \sum_{i=1}^{N^{i}} N^{i}(\xi) \mathbf{u}_{i} \quad T_{\mathbf{u}} V_{h}^{\mathbb{R}^{n}}(\Omega) = V_{h}^{\mathbb{R}^{n}}(\Omega) \subset H^{1}(\Omega, \mathbb{R}^{n})$$

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(B) Second-order, vertex degree of freedom



(c) Second-order, edge degree of freedom SANDER O (2016)

Outline





Open questions/Outlook



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Literature



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

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



Baustatik und Baudynamik



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Riemannian submanifolds

 $f(\mathbf{x}): \mathcal{M} \to \mathbb{R} \qquad \mathcal{M} \subset \mathbb{R}^n$

Metric of the embedding \bar{g}_x induces metric g_x on \mathcal{M}

 $g_{\mathbf{x}}(\boldsymbol{\xi}, \boldsymbol{\zeta}) = \bar{g}_{\mathbf{x}}(\boldsymbol{\xi}, \boldsymbol{\zeta}), \quad \boldsymbol{\xi}, \boldsymbol{\zeta} \in T_{\mathbf{x}} \mathcal{M}$





Spherical coordinates [edit]

We may define a coordinate system in an *n*-dimensional Euclidean space which is ar of a radial coordinate *r*, and *n* – 1 angular coordinates $\varphi_1, \varphi_2, \ldots, \varphi_{n-1}$, where the ang [0,360) degrees). If x_i are the Cartesian coordinates, then we may compute x_1, \ldots, x_n

 $\begin{array}{l} x_1 = r\cos(\varphi_1) \\ x_2 = r\sin(\varphi_1)\cos(\varphi_2) \\ x_3 = r\sin(\varphi_1)\sin(\varphi_2)\cos(\varphi_3) \\ \vdots \\ x_{n-1} = r\sin(\varphi_1)\cdots\sin(\varphi_{n-2})\cos(\varphi_{n-1}) \\ x_n = r\sin(\varphi_1)\cdots\sin(\varphi_{n-2})\sin(\varphi_{n-1}). \end{array}$

WIKIPEDIA: N-SPHERE



Riemannian gradient



Riemannian Gradient: submanifolds

$$f(\mathbf{x}): \mathcal{M} \to \mathbb{R}$$
 $\overline{f}(\mathbf{x}): \mathbb{R}^n \to \mathbb{R}$

"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.

 $\operatorname{grad} f(\mathbf{x}) = P_{\mathbf{x}} \operatorname{grad} \overline{f}(\mathbf{x})$



- No parametrization
- No artificial singularities
- Simple linearization



Toy problem, gradient and Riemannian gradient

Euclidean gradient Riemannian gradient



Riemannian Hessian



Riemannian Hessian: submanifolds

 $f(\mathbf{x}): \mathcal{M} \to \mathbb{R}$ $\overline{f}(\mathbf{x}): \mathbb{R}^n \to \mathbb{R}$

Levi-Civita connection:

 $abla_{\eta_{\mathbf{x}}} \boldsymbol{\xi} = P_{\mathbf{x}} \overline{
abla}_{\eta_{\mathbf{x}}} \boldsymbol{\xi} = P_{\mathbf{x}} \mathrm{D}_{\eta_{\mathbf{x}}} \boldsymbol{\xi} \qquad \boldsymbol{\xi}, \boldsymbol{\eta} \in T_{\mathbf{x}} \mathcal{M}$

 $\operatorname{Hess} f(\mathbf{x})\boldsymbol{\eta} = P_{\mathbf{x}} \operatorname{Hess} \overline{f}(\mathbf{x}) P_{\mathbf{x}} \boldsymbol{\eta} + W_{\mathbf{x}}(\boldsymbol{\eta}, P_{\mathbf{x}}^{\perp} \operatorname{grad} \overline{f}(\mathbf{x}))$

"[...] This shows that, for Riemannian submanifolds of Euclidean spaces, the Riemannian Hessian is the projected Euclidean Hessian plus a correction term which depends only on the normal part of the Euclidean gradient."

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

Algebraic optimization on manifolds



Reduction of dimensions

Gradient and Hessian, example unit sphere S^{n-1}

$$P_{\mathbf{x}} = \mathbf{I} - \mathbf{x} \otimes \mathbf{x}$$

$$\operatorname{grad} \Pi(\mathbf{x}) = P_{\mathbf{x}} \operatorname{grad} \bar{\Pi}(\mathbf{x}) \qquad \operatorname{Hess} \Pi(\mathbf{x}) = P_{\mathbf{x}} \operatorname{Hess} \bar{\Pi}(\mathbf{x}) - (\mathbf{x}^{T} \operatorname{grad} \bar{\Pi}(\mathbf{x}))\mathbf{I}$$

$$\operatorname{grad} \Pi(\mathbf{x}) \in T_{\mathbf{x}} \mathcal{S}^{n-1} \qquad \operatorname{Hess} \Pi(\mathbf{x}) \in (T_{\mathbf{x}} \mathcal{S}^{n-1} \times T_{\mathbf{x}} \mathcal{S}^{n-1})$$

Define a basis of the tangent space in S^2 at \mathbf{x}

$$\mathbf{\Lambda} = [\mathbf{x}_{I}^{1} \ \mathbf{x}_{I}^{2}] \in \mathbb{R}^{3 \times 2}$$

grad^{red} $\Pi(\mathbf{x})_{2 \times 1} = \mathbf{\Lambda}^{T}(P_{\mathbf{x}} \operatorname{grad} \bar{\Pi}(\mathbf{x})) = \mathbf{\Lambda}^{T}[\operatorname{grad} \bar{\Pi}(\mathbf{x})]_{3 \times 1}$
Hess^{red} $\Pi(\mathbf{x})_{2 \times 2} = \mathbf{\Lambda}^{T}[\operatorname{Hess} \bar{\Pi}(\mathbf{x})]_{3 \times 3}\mathbf{\Lambda}$

Resulting optimization algo has only n-1 dofs (as the dim. of the manifold)



Update of nodal values

Update of nodal values

 $\mathbf{x}_k + \Delta \mathbf{x}_k \not\in \mathcal{M}$



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

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

Along these geodesics one could perform e.g. gradient descent

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Algebraic optimization on manifolds



Update of nodal values

Absil PA, "Optimization on manifolds: methods and applications", Leuven, 18 Sep 2009. Luenberger, D.G. (1973) Introduction to Linear and Nonlinear Programming. Addison-Wesley, Boston. **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

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



Algebraic optimization on manifolds

Update of nodal values

Retractions for the unit sphere





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

Algebraic optimization on manifolds



Update of nodal values

Retractions for the unit sphere





Riemannian Newton

	Classic Newton	Riemannian Newton
Update	$\mathbf{x}_{k+1} = \mathbf{x}_k + \Delta \mathbf{x}_k$	$\mathbf{x}_{k+1} = R_{\mathbf{x}_k}(\Delta \mathbf{x}_k)$
Gradient	$\operatorname{grad} f(\mathbf{x})$	$\operatorname{grad} f(\mathbf{x}) = P_{\mathbf{x}} \operatorname{grad} \overline{f}(\mathbf{x})$
Hessian	$\operatorname{Hess} f(\mathbf{x})$	$\operatorname{Hess} f(\mathbf{x})\boldsymbol{\eta} = P_{\mathbf{x}} \operatorname{Hess} \overline{f}(\mathbf{x}) P_{\mathbf{x}} \boldsymbol{\eta} + W_{\mathbf{x}}(\boldsymbol{\eta}, P_{\mathbf{x}}^{\perp} \operatorname{grad} \overline{f}(\mathbf{x}))$

Ingredients:

 $\begin{array}{ll} R_{\mathbf{x}}: T_{\mathbf{x}}\mathcal{M} \to \mathcal{M} & \text{Hess } \bar{E}(\mathbf{x}) \\ \mathbf{P}_{\mathbf{x}}: \mathbb{R}^{n} \to T_{\mathbf{x}}\mathcal{M} & \text{grad } \bar{E}(\mathbf{x}) \\ W_{\mathbf{x}}: T_{\mathbf{x}}\mathcal{M} \times T_{\mathbf{x}}^{\perp}\mathcal{M} \to T_{\mathbf{x}}\mathcal{M} \end{array}$ Tangent space basis $\mathbf{\Lambda}$

Algebraic optimization on manifolds



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





Summary

	LAM	Penalty	Coords	Manifold optimization
Linearization	\odot	\odot	$\overline{\mathbf{S}}$	\odot
Singularities	\bigcirc	\odot	\odot	\odot
Search space dimensions	3	2	1	1
Minimization	$\overline{\mathbf{\cdot}}$	\odot	\odot	\odot
Iterations	\odot	$\overline{\mathbf{\cdot}}$	\odot	\odot

Outline









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Reissner-Mindlin: Roll-up of clamped beam

• 16 Iterations of Newton's method to reach equilibrium







Reissner-Mindlin: L-shape





Reissner-Mindlin: Buckling of a sheared sheet



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



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Reissner-Mindlin: Buckling of a sheared sheet



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Reissner-Mindlin: Buckling of a sheared sheet





GEOMETRICALLY NON-LINEAR REISSNER-MINDLIN SHELL MODEL, AM, BISCHOFF (2022): A CONSISTENT FINITE ELEMENT FORMULATION OF THE 00

Other physical problems



Simulation of micromagnetics



Maxwell's equation in vacuum and matter

Other physical problems



Simulation of micromagnetics



Open questions



Other manifolds

Manifold	Exponential map	Other retractions	Tangent space
Unit sphere S^{n-1}	$\cos \Delta \mathbf{x} \mathbf{x} + \frac{\sin \Delta \mathbf{x} }{ \Delta \mathbf{x} } \Delta \mathbf{x}$	$\frac{\mathbf{x} + \Delta \mathbf{x}}{ \mathbf{x} + \Delta \mathbf{x} }$	$T_{\mathbf{x}} \mathcal{S}^{n-1} = \{ \mathbf{y} \in \mathbb{R}^n \mid \mathbf{y}^T \mathbf{x} = 0 \}$
Special linear group $\mathcal{SL}(n) = \{ \mathbf{X} \in \mathbb{R}^{n \times n} \mid \det \mathbf{X} = 1 \}$	$\exp(\Delta \mathbf{X})\mathbf{X}$	$\frac{\mathbf{X} + \Delta \mathbf{X}}{\det(\mathbf{X} + \Delta \mathbf{X})^{1/n}}$	$T_{\mathbf{X}}\mathcal{SL}(n) = \{\mathbf{Y} \in \mathbb{R}^{n \times n} \mid \operatorname{tr} \mathbf{Y} = 0\}$
Symmetric Special linear manifold $\mathcal{SSL}(n) = \{ \mathbf{X} \in \mathbb{R}^{n \times n} \mid $ $\det \mathbf{X} = 1 \land \mathbf{X}^T = \mathbf{X} \}$	$\exp(\Delta \mathbf{X})\mathbf{X}$	$\frac{\mathbf{X} + \Delta \mathbf{X}}{\det(\mathbf{X} + \Delta \mathbf{X})^{1/n}}$	$T_{\mathbf{X}} \mathcal{SSL}(n) = \{ \mathbf{Y} \in \mathbb{R}^{n \times n} \mid \operatorname{tr}(\mathbf{X}^{-1}\mathbf{Y}) = 0 \}$
Special orthogonal group $\mathcal{SO}(n) = \{ \mathbf{X} \in \mathbb{R}^{n \times n} \mid $ $\mathbf{X}^T \mathbf{X} = \mathbf{I} \land \det \mathbf{X} = 1 \}$	$\exp(\Delta \mathbf{X})\mathbf{X}$	QR decomposition	$T_{\mathbf{X}}\mathcal{SO}(n) = \{\mathbf{Y} \in \mathbb{R}^{n \times n} \mathbf{Y}^T = -\mathbf{Y}\}$

Outlook



Apply results

• Geometrically nonlinear beam

Dynamics on manifolds

- Non-constant mass matrix
- Riemannian Hamiltonian/Lagrangian
- Variational integrators for manifolds

• ...

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Thank you!



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