

Model reduction for geometrically nonlinear elasticity

Motivation:

Physical experiments show that the frequently used linear model for elastic deformation problems is only valid up to a certain strain. The usage of nonlinear models coincide also for larger strains better with these experimental results and should hence be used in numerical simulations. But also here a variety of nonlinear models are possible, e.g. a Neo-Hooke model, a more general Mooney-Rivlin model or even a more complex one.

Moreover, solving a nonlinear model needs in general more effort than solving a linear model, e.g. due to linearization techniques. The idea of model adaptivity is to use a nonlinear model only in regions of the domain where it is necessary and use the linear model on the other part.

Reduced mixed least squares method:

For the implementation of a model adaptivity technique we use a mixed least squares finite element method and choose a Neo-Hooke model with stored energy function

$$\psi_{NH}(\mathbf{C}) = \frac{\mu}{2} \text{tr}(\mathbf{C}) + \frac{\lambda}{4} \det \mathbf{C} - \left(\frac{\mu}{2} + \frac{\lambda}{4} \right) \ln(\det \mathbf{C})$$

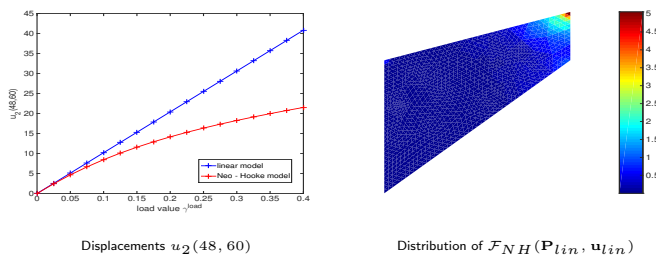
as nonlinear model.

We decompose a given body $\Omega \subset \mathbb{R}^n$ into a „nonlinear“ subdomain $\Omega_1 \subseteq \Omega$ and a „linear“ subdomain $\Omega_2 = \Omega \setminus \Omega_1$. For such an decomposition of Ω we seek a minimizer of a least squares functional of the form

$$\mathcal{F}_{red}(\mathbf{P}, \mathbf{u}) = \|\mathcal{R}_{NH}(\mathbf{P}, \mathbf{u})\|_{L^2(\Omega_1)}^2 + \|\mathcal{R}_{lin}(\mathbf{P}, \mathbf{u})\|_{L^2(\Omega_2)}^2. \quad (1)$$

Essential steps in the algorithm:

- Start with a full linear model in step $i = 0$, i. e. $\Omega_1^{(0)} = \emptyset$, $\Omega_2^{(0)} = \Omega$ and solve (1) with solution $(\mathbf{P}_{red}^{(0)}, \mathbf{u}_{red}^{(0)}) = (\mathbf{P}_{lin}, \mathbf{u}_{lin})$.
- Adapt the nonlinear subdomain by choosing $\Omega_1^{(i+1)}$ with the help of the nonlinear least squares functional $\mathcal{F}_{NH}(\mathbf{P}_{red}^{(i)}, \mathbf{u}_{red}^{(i)})$ and a suitable marking strategy. Set $\Omega_2^{(i+1)} = \Omega \setminus \Omega_1^{(i+1)}$ and determine the solution $(\mathbf{P}_{red}^{(i+1)}, \mathbf{u}_{red}^{(i+1)})$ of (1) on the new decomposition, set $i = i + 1$ and continue in the same manner.
- Nonlinear elements in previous step remain nonlinear.
- Use Gauss-Newton method for nonlinear part in (1).

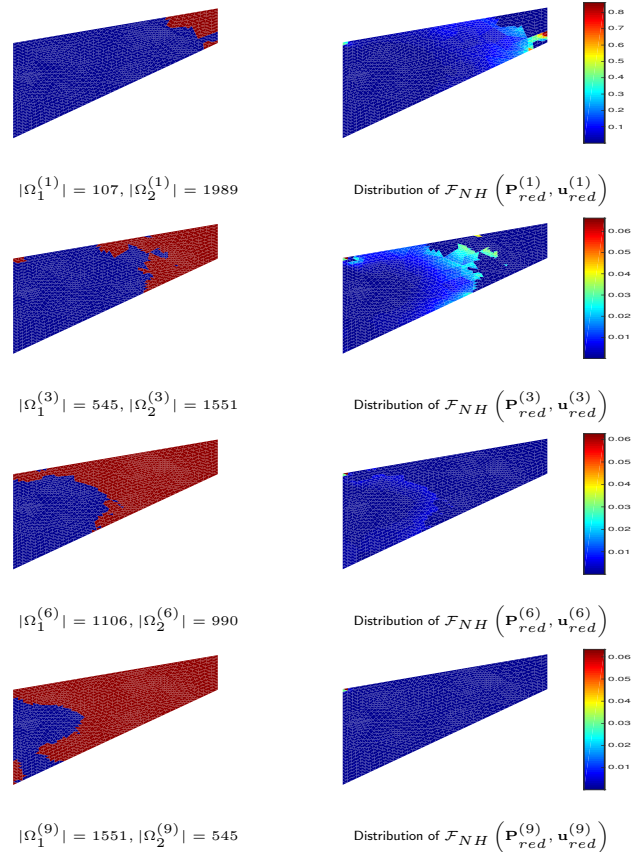


Numerical results for model adaptivity (Cook's membrane in 2d)

Given:

$\mathbf{u} = \mathbf{0}$ (left boundary), $\mathbf{P} \cdot \mathbf{n} = \mathbf{0}$ (top/bottom boundary) and $\mathbf{P} \cdot \mathbf{n} = (0, \gamma^{\text{load}})^T$ (right boundary) with load parameter $\gamma^{\text{load}} = 0.25$, Lamé constants: $\lambda = \infty$, $\mu = 1$

Below the decomposition of the domain Ω into the linear part $\Omega_2^{(i)}$ and the nonlinear part $\Omega_1^{(i)}$ (left) and the distribution $\mathcal{F}_{NH}(\mathbf{P}_{red}^{(i)}, \mathbf{u}_{red}^{(i)})$ (right) is depicted:



Benefits:

- Model adaptivity techniques, also in the context of LSFEM, may help for speeding up an algorithm.
- Quadrature error for the reduced model in comparison to the nonlinear model becomes smaller.
- Combining model adaptivity and usual mesh refinement is possible.

References:

- [1] B. Müller, G. Starke, A. Schwarz, J. Schröder: A First-Order System Least Squares Method for Hyperelasticity. SIAM J. Sci. Comput. 36(5): B795-B816, 2014
- [2] B. Müller: Mixed Least Squares Finite Element Methods Based on Inverse Stress-Strain Relations in Hyperelasticity. Ph.D. Thesis, Universität Duisburg-Essen, 2014.