

Recent applications of Ferrite.jl at the IKM



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Overview

- complex material models in Ferrite.jl
 - time-separated **stochastic** mechanics → Geisler
 - **damage** and fatigue modelling → Kök
 - topology **optimization** with plasticity → Kick
 - topology **optimization** for large deformations → von Zabiensky



Uncertainty quantification for inelastic material models

Up to 1,200 Airbus Jet Engines Recalled

The recall was caused by on impurities in powdered metal.



BY SÉBASTIEN ROBLIN PUBLISHED: JUL 27, 2023



source: prattwhitney.com

BUSINESS

Pratt & Whitney Engines on Hundreds of Airbus Jets Recalled for Inspection

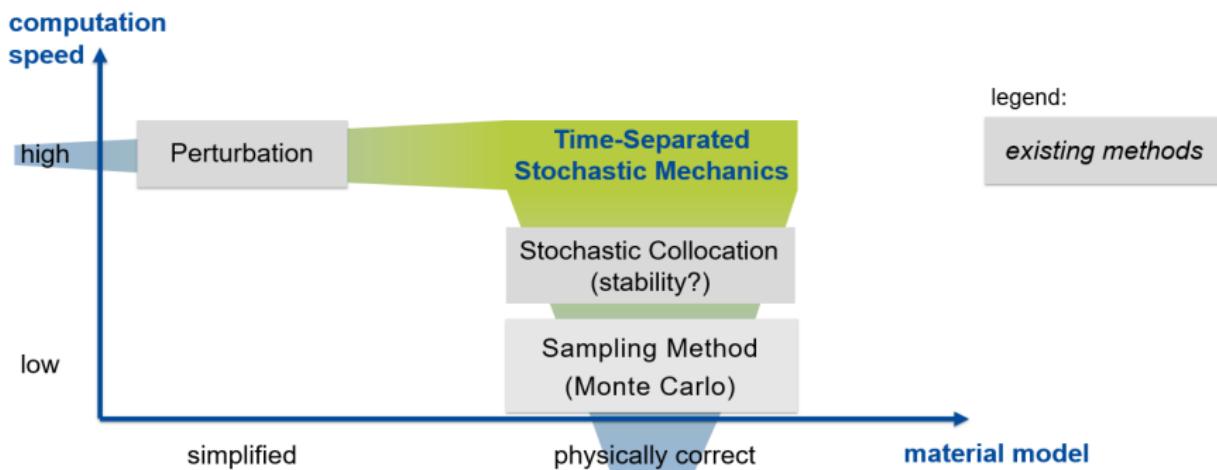
RTX says engines are affected by contaminated metal parts that could crack over time



source: prattwhitney.com

How to **predict** the stochastic behavior **efficiently and accurately?**

State-of-the-art





Implementation in Ferrite

Approach: split into **deterministic** "0" and **stochastic** part "I":

$$\rightarrow \mathbb{E} = \mathbb{E}^0 + \xi \mathbb{E}^I, \langle \xi \rangle = 0$$

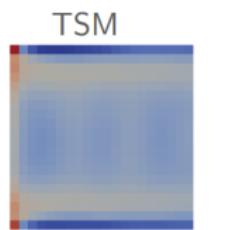
$$\rightarrow \varepsilon = \varepsilon^0 + \xi \varepsilon^I, \quad \varepsilon^{inelas} = \varepsilon^{inelas,0} + \xi \varepsilon^{inelas,I}$$

Separate set of PDEs for deterministic and stochastic part

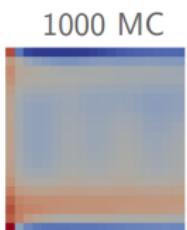
→ Evolution equations stem naturally from standard set of equations

- 1: **for** each loadstep **do**
- 2: **do FEM** "0" $\rightarrow \varepsilon^0, \varepsilon^{inelas,0}$
- 3: **do FEM** "I" $\rightarrow \varepsilon^I, \varepsilon^{inelas,I}$
- 4: **calculate expectation & standard deviation**
 of stresses, reaction forces, etc.
- 5: **end for**

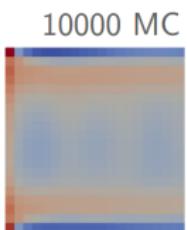
Results for standard deviation of stress



0min 17s

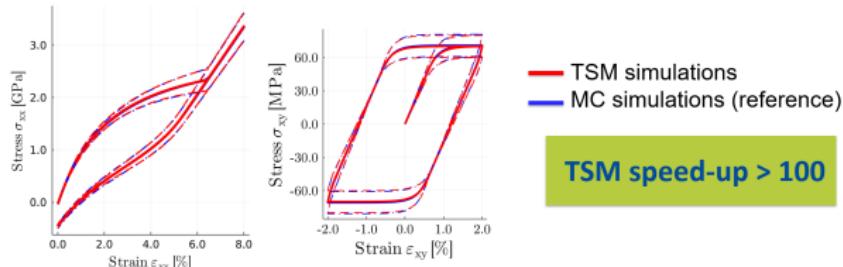
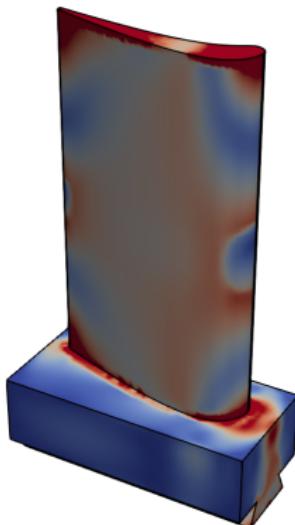


18min 40s



3h 3min

18h (TSM) vs.
0.7 years (MC)



→ viscoplasticity, damage, phasetransformations ...



Hamilton-based damage model (HDM)

- extended Hamilton functional e.g. for damage modelling

$$\mathcal{H}[\boldsymbol{u}, \boldsymbol{\alpha}] := \underbrace{\mathcal{G}[\boldsymbol{u}, \boldsymbol{\alpha}]}_{\text{Gibbs energy}} + \underbrace{\mathcal{D}[\boldsymbol{\alpha}]}_{\text{dissipation}} - \underbrace{\mathcal{R}[\boldsymbol{\alpha}]}_{\text{hardening}} + \underbrace{\mathcal{C}[\boldsymbol{\alpha}]}_{\text{constraints}}$$

- Internal variables: $\boldsymbol{\alpha} = \{d, \boldsymbol{\varepsilon}_p\}$

$$\hookrightarrow \Psi(\boldsymbol{\varepsilon}, \boldsymbol{\alpha}) = \frac{1}{2}(\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_p) : f(d)\mathbb{C} : (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_p) + \frac{1}{2}\beta \|\nabla f\|^2$$

$$\hookrightarrow \mathcal{D}[\boldsymbol{\alpha}] = \int_{\Omega} \frac{\partial \Delta}{\partial \dot{\boldsymbol{\alpha}}} \cdot \boldsymbol{\alpha} \, dV$$

$$\hookrightarrow \Delta[\boldsymbol{\alpha}] = r_d |\dot{d}| + \frac{1}{2} \eta_d |\dot{d}|^2 + r_p \|\dot{\boldsymbol{\varepsilon}}_p\| + \frac{1}{2} \eta_p \|\dot{\boldsymbol{\varepsilon}}_p\|^2$$

$$\hookrightarrow \mathcal{R}[\boldsymbol{\alpha}] = \int_{\Omega} \Psi_h \, dV \text{ with } \Psi_h = \Psi_h(\boldsymbol{\alpha}).$$

$$\hookrightarrow \mathcal{C}[\boldsymbol{\alpha}] = \int_{\Omega} p_c^p : \boldsymbol{\varepsilon}_p + p_c^\alpha \, dV$$



Hamilton-based damage modelling (HDM)

- requiring stationarity: $\mathcal{H}[\mathbf{u}, \alpha, \cdot] \rightarrow_{\mathbf{u}, \alpha}^{\text{stat}}$
- stationarity condition yields:

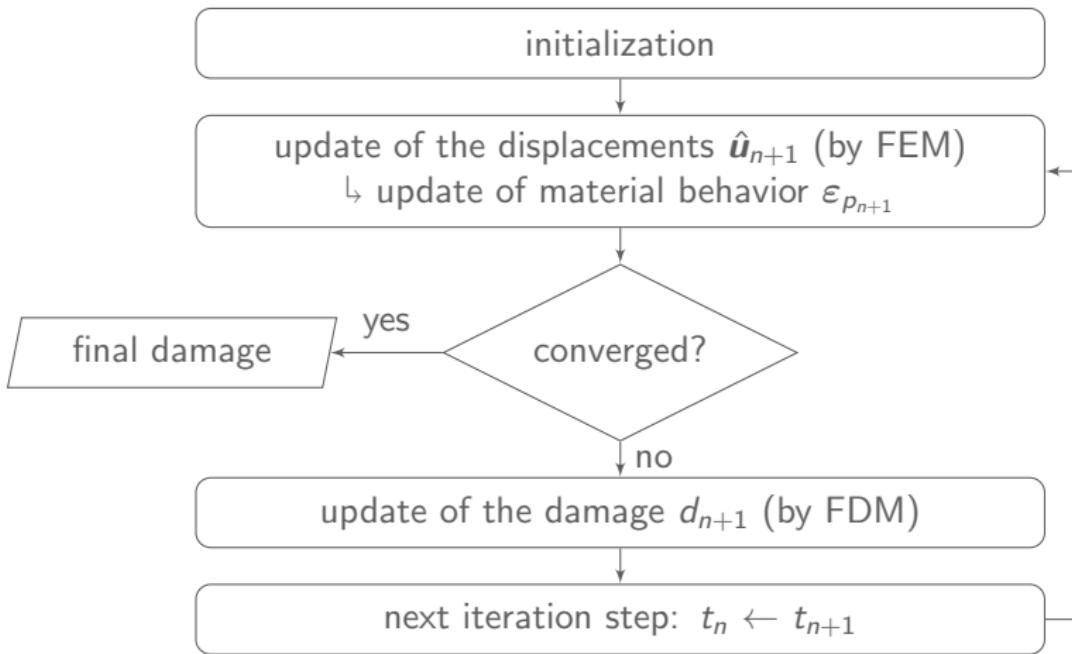
$$\begin{cases} \int_{\Omega} \frac{\partial \Psi_m}{\partial \varepsilon} \delta \mathbf{u} \, dV - \int_{\Omega} \mathbf{b}^* \cdot \mathbf{u} \delta \mathbf{u} \, dV - \int_{\partial\Omega} \mathbf{t}^* \cdot \mathbf{u} \delta \mathbf{u} \, dA &= 0 \quad \forall \delta \mathbf{u} \\ \int_{\Omega} (-\mathbb{C} : (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_p) + \frac{\partial \Delta}{\partial \boldsymbol{\varepsilon}_p} + \kappa \mathbf{I}) : \delta \boldsymbol{\varepsilon}_p \, dV &= 0 \quad \forall \delta \boldsymbol{\varepsilon}_p \\ \int_{\Omega} \Psi_0 f' \delta d \, dV - \int_{\Omega} \beta \nabla f \cdot \nabla (f' \delta d) \, dV + \int_{\Omega} \frac{\partial \Delta}{\partial d} \, dA \delta d &= 0 \quad \forall \delta d \end{cases}$$



Hamilton-based damage modelling (HDM)

- solve system of stationary conditions
 - $\delta_u \mathcal{H} = 0 \Rightarrow$ weak form of balance of linear momentum
 - $\delta_{\epsilon_p} \mathcal{H} = 0 \Rightarrow$ material model (evolution equation)
 - ⇒ solved by **finite element method** (FEM)
 - $\delta_d \mathcal{H} = 0 \Rightarrow$ damage equation (strong form)
$$\Psi_0 f(d) - \beta f(d) \nabla^2 f(d) - r - \eta \dot{d} \leq 0$$
 - ⇒ solved by **finite difference method** (FDM)
- staggered FEM + FDM ⇒ **NEM** (neighbored element method)

Damage Process





Algorithm: update of \mathbf{u}_{n+1} and $\boldsymbol{\varepsilon}_{pn+1}$ by FEM

```
1: while true do
2:   for each element  $\in$  mesh do
3:     call reinit!(mesh, elementvalues)
4:     call  $K_e, \mathbf{r}_e = \text{assembleCell}!(\text{elementvalues}, \mathbf{u})$             $\triangleright$  see next slide
5:     call assemble!(assembler,  $K_e, \mathbf{r}_e$ )
6:   end for
7:   call apply_zero!( $K, \mathbf{r}, \text{constraints}$ )
8:   if  $\|\mathbf{r}\| < \text{tol}$  then break
9:   end if
10:  update  $\mathbf{u}_{i+1} = \mathbf{u}_i - K^{-1}\mathbf{r}$ 
11:  update  $i = i + 1$ 
12: end while
13:  $\mathbf{u}_{n+1} \leftarrow \mathbf{u}_{i+1}$ 
```



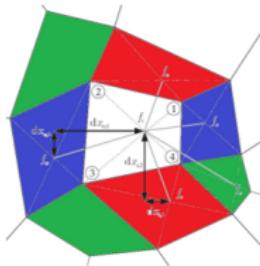
Algorithm: update of \mathbf{u}_{n+1} and $\boldsymbol{\varepsilon}_{pn+1}$ by FEM

- assembleCell!(*elementvalues*, \mathbf{u})

```
1: for each  $ip \in element$  do
2:   compute  $\boldsymbol{\varepsilon}_{n+1} = \text{function\_symmetric\_gradient}(\text{elementvalues}, ip, \mathbf{u}_e)$ 
3:   compute  $\boldsymbol{\sigma}_{n+1}(\mathbf{d}_n)$ ,  $\mathbb{D}_{0,n+1}$ ,  $\boldsymbol{\varepsilon}_{pn+1}$  and  $\Psi_{0,n+1} \rightarrow$  material model
4:   compute  $\Omega^* = \text{getdetJdV}(\text{elementvalues}, ip)$ 
5:   for  $i$  to number base shape functions do
6:     compute  $\mathbf{B}^T = \text{shape\_symmetric\_gradient}(\text{elementvalues}, ip, i)$ 
7:     compute  $\mathbf{r}_e[i] += (\mathbf{B}^T : \boldsymbol{\sigma}_{n+1}) \Omega^*$ 
8:     for  $j$  to number base shape functions do
9:       compute  $\mathbf{B} = \text{shape\_symmetric\_gradient}(\text{elementvalues}, ip, j)$ 
10:      compute  $\mathbf{K}_e[i, j] += (\mathbf{B}^T : \mathbb{D}_{0,n+1} : \mathbf{B}) \Omega^*$ 
11:    end for
12:  end for
13:  compute  $\mathbf{r}_e = \mathbf{r}_e - \mathbf{f}_{e,\text{ext}}$ 
14: end for
```

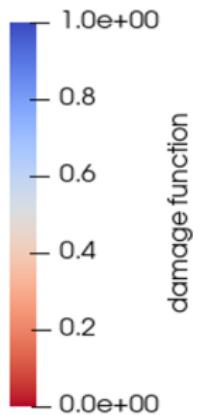
Update of d_{n+1} via "external" FDM

- input:
 - strain energy: $\Psi_{0,n+1}$ (from FEM)
 - structure volume (user input)
 - regularization parameter (user input) $\beta \rightarrow$ minimum member size
- compute
 - Laplace operator $\nabla^2 f(d)$
 - requires mesh data
 - new design d_{n+1} by solving PDE via FDM
 - within two nested loops independent of FEM





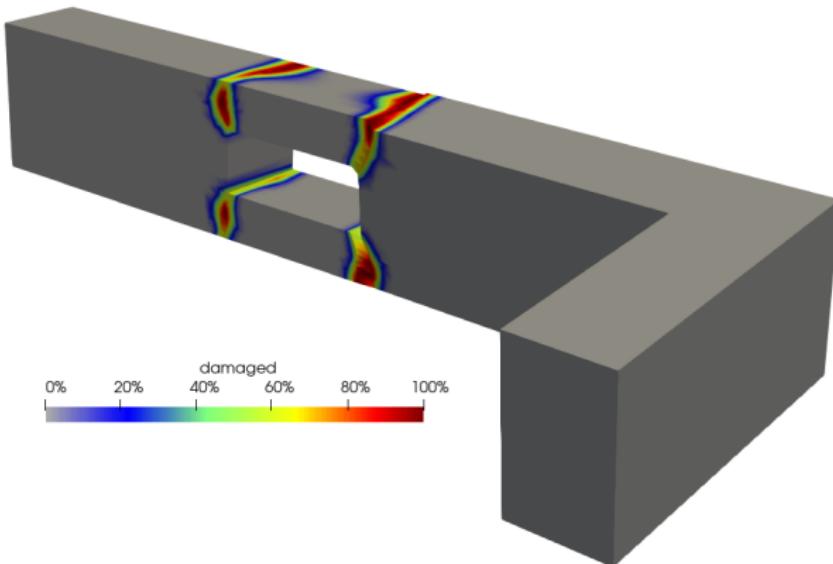
Damage simulation results



damage function



Damage simulation results





The potency of topology optimization



Thermodynamic topology optimization (TTO)

- extended Hamilton functional e.g. for topology optimization

$$\mathcal{H}[\mathbf{u}, \boldsymbol{\alpha}, \chi] := \underbrace{\mathcal{G}[\mathbf{u}, \boldsymbol{\alpha}, \chi]}_{\text{Gibbs energy}} + \underbrace{\mathcal{D}[\boldsymbol{\alpha}]}_{\text{dissipation}} - \underbrace{\mathcal{R}[\chi]}_{\text{rearrangement}} + \underbrace{\mathcal{C}[\boldsymbol{\alpha}, \chi]}_{\text{constraints}}$$

↳ **design variable(s)** χ besides internal variables $\boldsymbol{\alpha}$

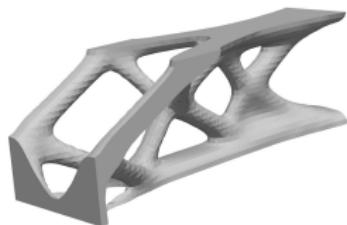
e.g. $\chi \rightarrow$ material distribution = topology

$$\left. \begin{array}{l} \mathcal{G} = \dots \\ \mathcal{D} = \dots \\ \mathcal{R} = \dots \\ \mathcal{C} = \dots \end{array} \right\} \Rightarrow \text{Already well documented example in Ferrite.jl! :-)}$$

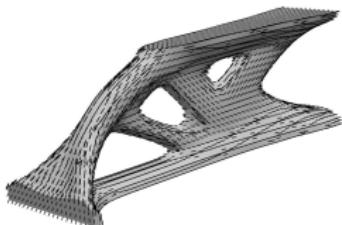
↳ Solution via NEM analogous to damage model

Different elastaic materials

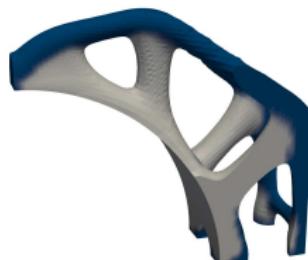
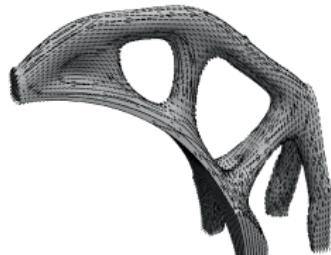
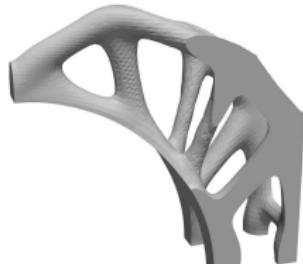
linear-elastic



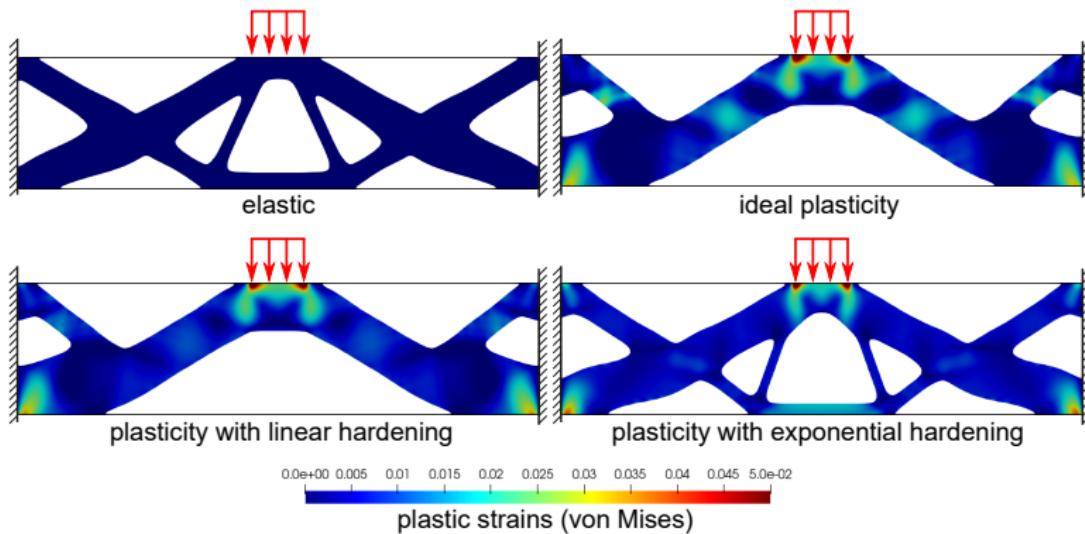
anisotropic



mixed material



Plasticity with different kinds of hardening



Implementation of large deformations in TTO

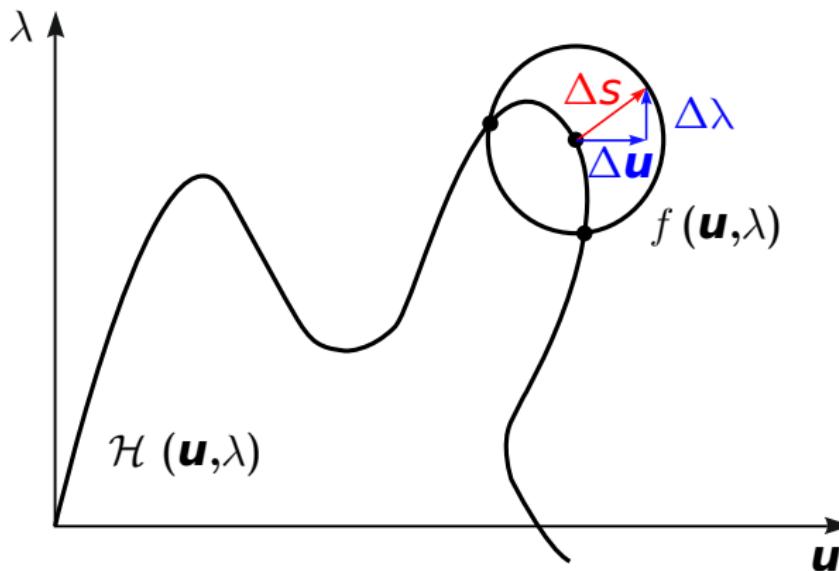
- if a **Hyperelastic Material Model** is used:
 - **large** displacements \mathbf{u} and strains $\boldsymbol{\varepsilon}$
 - $\delta_{\mathbf{u}} \mathcal{H} = 0$ is **non-linear**
 - **snap-back** and **snap-through** possible
 - geometrical part of B-operator necessary
- discretized system for **arc-length method**
 - $$\begin{pmatrix} \mathbf{K} & -\mathbf{P} \\ \mathbf{f}^T & f_{,\lambda} \end{pmatrix}_i \begin{Bmatrix} \Delta \mathbf{u} \\ \Delta \lambda \end{Bmatrix}_{i+1} = \begin{Bmatrix} \mathcal{H} \\ \mathbf{f} \end{Bmatrix}_i$$
- with an additional constrain $f(\mathbf{u}, \lambda)$
 - $f = \sqrt{(\Delta \mathbf{u})^T (\Delta \mathbf{u}) + (\Delta \lambda)^2} - \Delta s$



<https://3druck.com/forschung/harvard-forscher-drucken-softroboter-mit-integrierten-sensoren-5968339/>

Visualization of the arc-length method

- solve problems with snap-back and snap-through





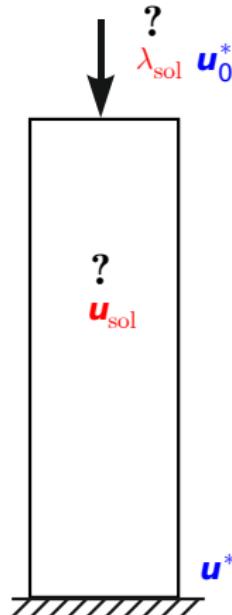
Algorithm: arc-length method according to [1]

```
1: set  $\mathbf{u}_0 = \mathbf{u}_k$  and  $\Delta s$                                 ▷ First Settings by User
2: compute  $\mathbf{u}_{P_0} = \mathbf{P}/\mathbf{K}_0$                          ▷ Predictor Step
3: compute  $\lambda_0 = \lambda_k \pm \frac{\Delta s}{\sqrt{(\Delta \mathbf{u}_{P_0})^T \Delta \mathbf{u}_{P_0}}}$ 
4: while  $\|\mathcal{H}(\mathbf{u}, \lambda)\| \leq \text{TOL}$  do
5:   for  $i = 0, 1, 2, \dots$  do
6:     compute  $\Delta \mathbf{u}_{P_{i+1}} = \mathbf{P}/\mathbf{K}_i$ 
7:     compute  $\Delta \mathbf{u}_{\mathcal{H}_{i+1}} = -\mathcal{H}(\mathbf{u}_i, \lambda_i)/\mathbf{K}_i$ 
8:     compute  $\Delta \lambda_{i+1} = \frac{f_i + f_i^T \Delta \mathbf{u}_{\mathcal{H}_{i+1}}}{f_i \lambda_i + f_i^T \Delta \mathbf{u}_{P_{i+1}}}$            ▷ Increments for next Step
9:     compute  $\Delta \mathbf{u}_{i+1} = \Delta \lambda_{i+1} \Delta \mathbf{u}_{P_{i+1}} + \Delta \mathbf{u}_{\mathcal{H}_{i+1}}$ 
10:    update  $\lambda_{i+1} = \lambda_i + \Delta \lambda_{i+1}$ 
11:    update  $\mathbf{u}_{i+1} = \mathbf{u}_i + \Delta \mathbf{u}_{i+1}$ 
12:    update  $\mathcal{H}(\mathbf{u}_{i+1}, \lambda_{i+1})$  and  $i = i + 1$ 
13:  end for
14: end while
```

A new extended solution method according to [2]

- boundary conditions based on
displacements only

$$\bullet \quad \boldsymbol{u} = \begin{cases} \quad \quad \quad \boldsymbol{u}_{\text{sol}} \\ \quad \quad \quad \boldsymbol{u}^* \\ \lambda_{\text{sol}} \quad \boldsymbol{u}_0^* \end{cases}$$



- how to implement \boldsymbol{u} in DoF-Handler?



Conclusions

- Developed at our institute in Ferrite.jl
 - **staggered models**
 - not "simply" coupled FEM problems
 - **external algorithms for material update**
 - "add-ons" to FEM with **modifications to core**
- Wishlist and eager to discuss
 - large deformations
 - geometrical part of shape derivatives
 - **arc length method** incl. displacement based approach
 - 4D Space-Time → for the next FerriteCon :-)



References

- [1] Peter Wriggers. *Nichtlineare finite-element-methoden*. Springer-Verlag, 2013.
- [2] Giuliano Petti, William M. Coombs, and Charles E. Augarde. A displacement-controlled arc-length solution scheme. *Computers & Structures*, 258:106674, 2022.
- [3] Miriam Kick and Philipp Junker. Thermodynamic topology optimization including plasticity. *arXiv preprint arXiv:2103.03567*, 2021.
- [4] Hendrik Geisler, Jan Nagel, and Philipp Junker. Simulation of the dynamic behavior of viscoelastic structures with random material parameters using time-separated stochastic mechanics. *submitted*, 2022.
- [5] Dustin R. Jantos, Klaus Hackl, and Philipp Junker. An accurate and fast regularization approach to thermodynamic topology optimization. *International Journal for Numerical Methods in Engineering*, 117(9):991–1017, 2019.
- [6] Andreas Vogel and Philipp Junker. Adaptive and highly accurate numerical treatment for a gradient-enhanced brittle damage model. *International Journal for Numerical Methods in Engineering*, 121(14):3108–3131, July 2020.



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Thank you for your kind attention!



école
normale
supérieure
paris-saclay



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