

Recent applications of Ferrite.jl at the IKM



Dustin R. Jantos, Ilayda Kök,
Hendrik Geisler, Miriam Kick,
Max von Zabiensky

Institute of Continuum Mechanics,
Leibniz Universität Hannover

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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

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1 0 0 4



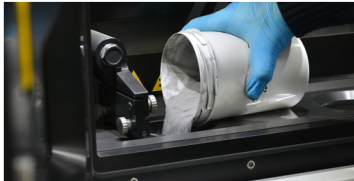
Uncertainty quantification for inelastic material models

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



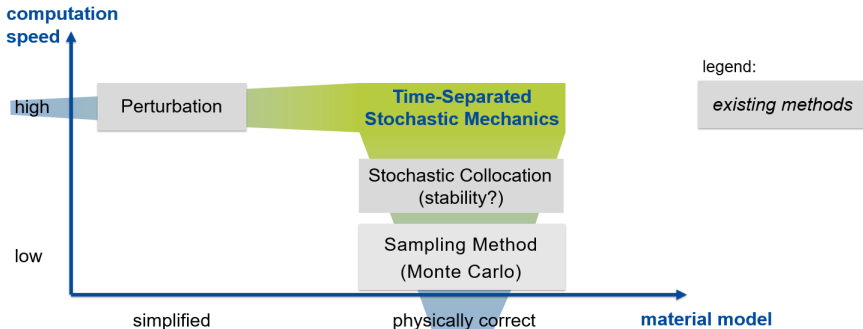
source: prattwhitney.com

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

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State-of-the-art





Implementation in Ferrite

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

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

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

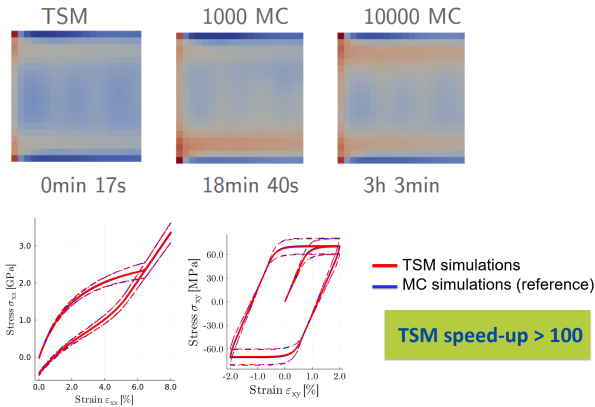
Seperate set of PDEs for deterministic and stochastic part

→ Evolution equations stem naturally from standard set of equations

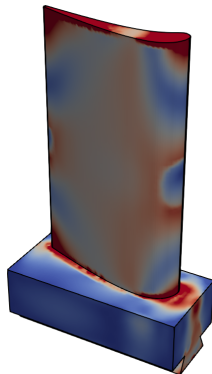
- 1: **for** each loadstep **do**
- 2: do FEM "0" → $\boldsymbol{\varepsilon}^0, \boldsymbol{\varepsilon}^{inelas,0}$
- 3: do FEM "I" → $\boldsymbol{\varepsilon}^I, \boldsymbol{\varepsilon}^{inelas,I}$
- 4: calculate expectation & standard deviation
 of stresses, reaction forces, etc.
- 5: **end for**



Results for standard deviation of stress



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



→ viscoplasticity, damage, phasetransformations ...



Hamilton-based damage model (HDM)

- extended Hamilton functional e.g. for damage modelling

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

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

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

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

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

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

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



Hamilton-based damage modelling (HDM)

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

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

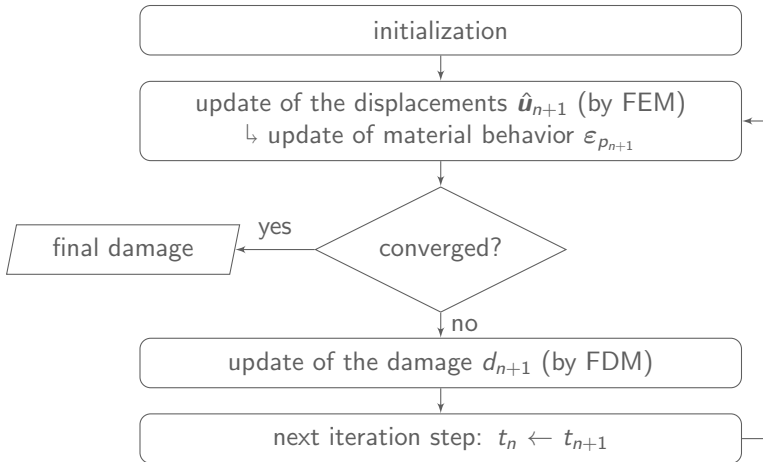


Hamilton-based damage modelling (HDM)

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



Damage Process





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

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



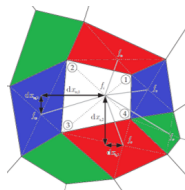
Algorithm: update of \mathbf{u}_{n+1} and $\boldsymbol{\varepsilon}_{pn+1}$ by FEM

- assembleCell!(*elementvalues*, \mathbf{u})
 - 1: **for each** $ip \in element$ **do**
 - 2: compute $\boldsymbol{\varepsilon}_{n+1} = \text{function_symmetric_gradient}(elementvalues, ip, \mathbf{u}_e)$
 - 3: compute $\boldsymbol{\sigma}_{n+1}(d_n)$, $\mathbb{D}_{0,n+1}$, $\boldsymbol{\varepsilon}_{pn+1}$ and $\boldsymbol{\Psi}_{0,n+1} \rightarrow$ **material model**
 - 4: compute $\Omega^* = \text{getdetJdV}(elementvalues, ip)$
 - 5: **for** i to number base shape functions **do**
 - 6: compute $\mathbf{B}^T = \text{shape_symmetric_gradient}(elementvalues, ip, i)$
 - 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 $\mathbf{B} = \text{shape_symmetric_gradient}(elementvalues, ip, j)$
 - 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 $\mathbf{r}_e = \mathbf{r}_e - \mathbf{f}_{e,ext}$
 - 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

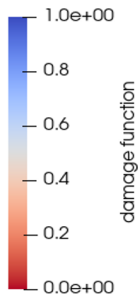
- compute
 - Laplace operator $\nabla^2 f(d)$
 - \rightarrow requires mesh data
 - new design d_{n+1} by solving PDE via FDM
 - \rightarrow within two nested loops independent of FEM



1 1
1 0 2
1 0 0 4

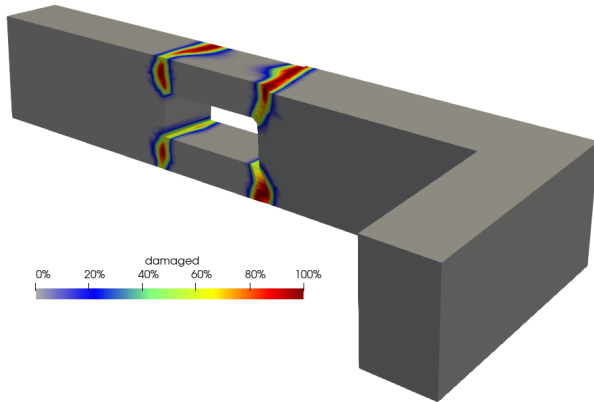


Damage simulation results





Damage simulation results





The potency of topology optimization

1 1
1 0 2
1 0 0 4



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

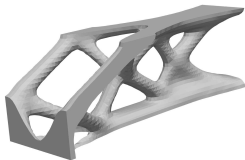
$$\left. \begin{array}{l} \mathcal{G} = \dots \\ \mathcal{D} = \dots \\ \mathcal{R} = \dots \\ \mathcal{C} = \dots \end{array} \right\} \Rightarrow \text{Already well documented} \\ \text{example in Ferrite.jl! :-)}$$

↳ Solution via NEM analogous to damage model

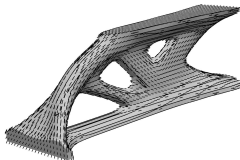


Different elastaic materials

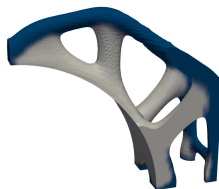
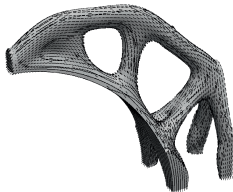
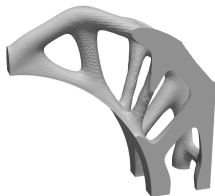
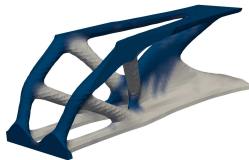
linear-elastic



anisotropic



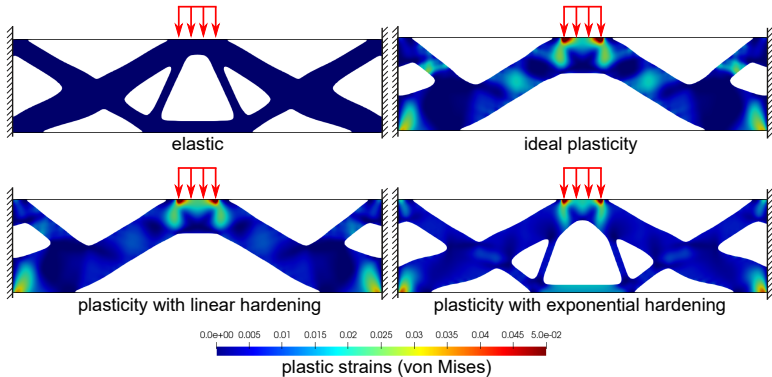
mixed material



1 1
1 0 2
0 0 4



Plasticity with different kinds of hardening



11
102
1004



Implementation of large deformations in TTD

- 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



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

- discretized system for **arc-length method**

$$\circ \begin{pmatrix} \mathbf{K} & -\mathbf{P} \\ f^T & f_{,\lambda} \end{pmatrix}_i \begin{Bmatrix} \Delta \mathbf{u} \\ \Delta \lambda \end{Bmatrix}_{i+1} = \begin{Bmatrix} \mathcal{H} \\ f \end{Bmatrix}_i$$

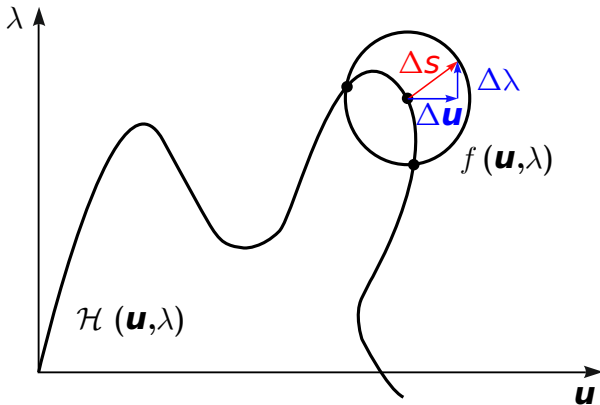
- with an additional constrain $f(\mathbf{u}, \lambda)$

$$\circ f = \sqrt{(\Delta \mathbf{u})^T (\Delta \mathbf{u}) + (\Delta \lambda)^2} - \Delta s$$

1 1
1 0 2
1 0 0 4

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 Δs ▷ 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 $\Delta \lambda_{i+1} = \frac{\mathbf{f}_i + \mathbf{f}_i^T \Delta \mathbf{u}_{\mathcal{H}_{i+1}}}{\mathbf{f}_i, \lambda_i + \mathbf{f}_i^T \Delta \mathbf{u}_{P_{i+1}}}$ ▷ 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**

$$\begin{array}{r} 11 \\ 102 \\ \hline 1004 \end{array}$$

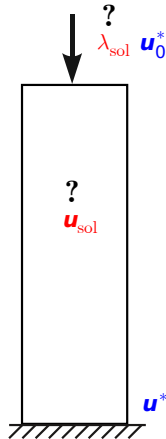


A new extended solution method according to [2]

- boundary conditions based on **displacements only**

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

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



$$\begin{array}{r} 11 \\ 102 \\ 1004 \end{array}$$



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.
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- [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.
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www.ikm.uni-hannover.de



@ikm.luh