

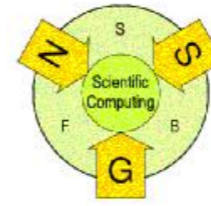
Multi-yield elastoplastic continuum - modeling and computations.

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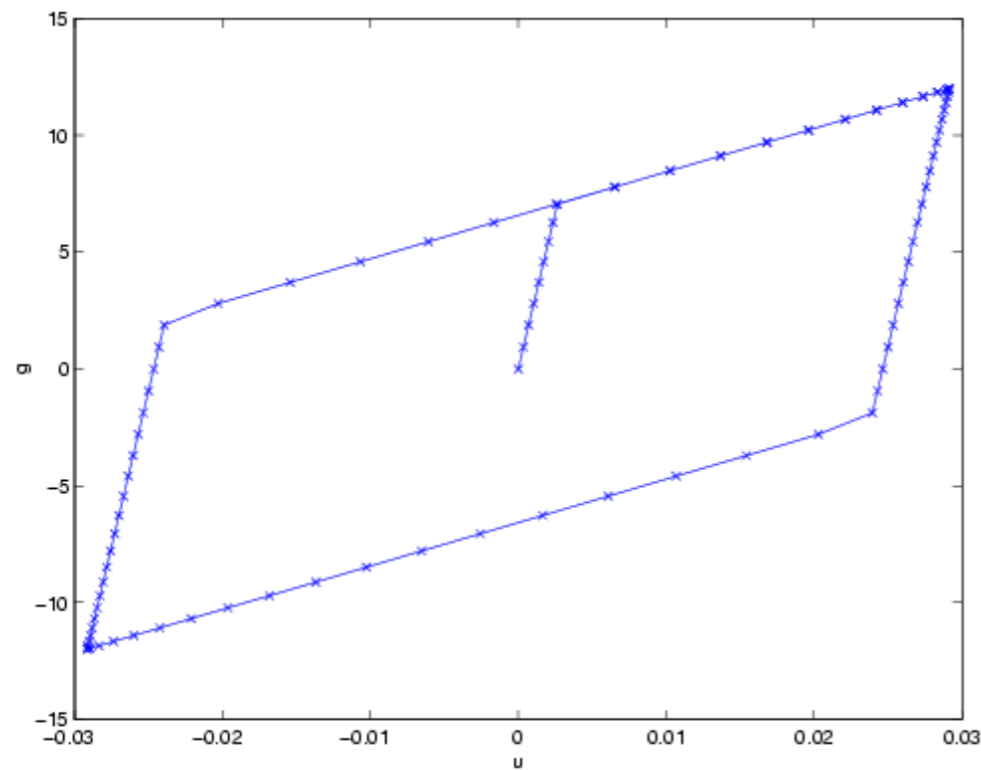


Outline

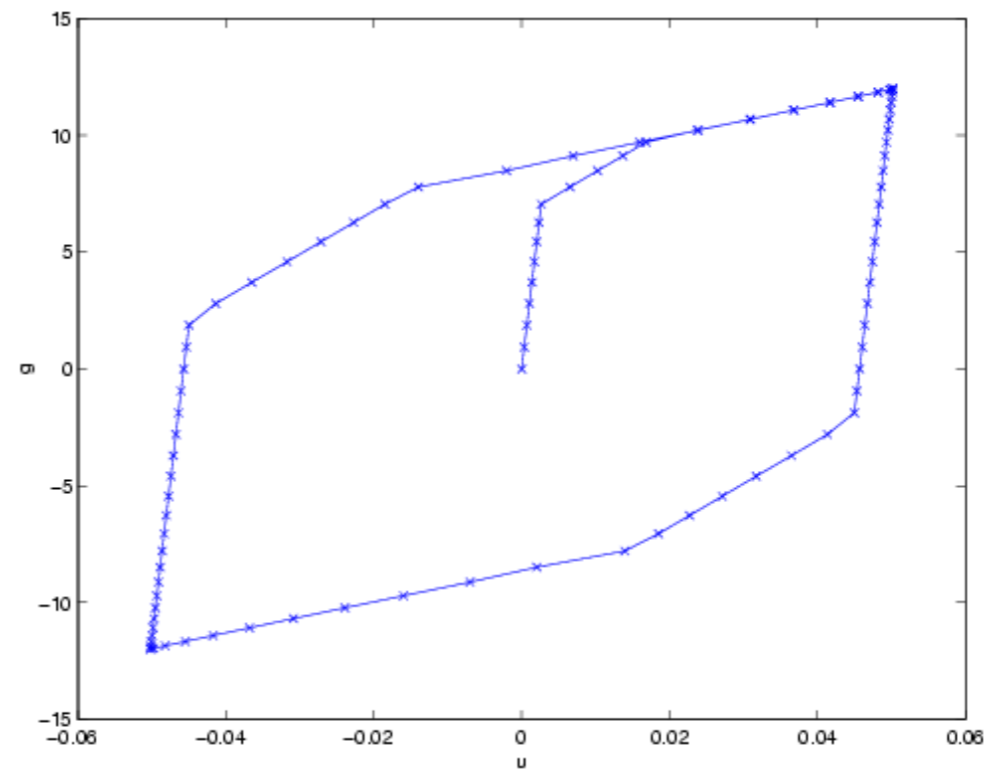
- Why multi-yield plasticity?
- Mathematical analysis
- Discretized minimization problem
- Algorithm
- Numerical experiments
- Outlook

Why Multi-yield (Two-yield) model?

- More realistic hysteresis stress-strain relation in materials!



Kinematic hardening model.



Two-yield hardening model.

Mathematical model of M-yield elastoplasticity

Problem (Prandtl-Ishlinskii): For $l \in H^1(0, T; \mathcal{H}^*)$, $l(0) = 0$, find $w = (u, p^1, \dots, p^M) : [0, T] \rightarrow \mathcal{H}$, $w(0) = 0$ s. t.

$$\langle l(t), z - \dot{w}(t) \rangle \leq a(w(t), z - \dot{w}(t)) + j(z) - j(\dot{w}(t))$$

$$\forall z = (v, \tau^1, \dots, \tau^M) \in \mathcal{H}.$$

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Notation: $\mathcal{H} = H_D^1(\Omega) \times \underbrace{L^2(\Omega)_{sym}^{d \times d} \times \dots \times L^2(\Omega)_{sym}^{d \times d}}_{M \text{ times}}$,

$$a(w, z) = \int_{\Omega} \left(\mathbb{C}(\varepsilon(u) - \sum_{r=1}^M p^r) \right) : \left(\varepsilon(v) - \sum_{r=1}^M \tau^r \right) dx + \sum_{r=1}^M \int_{\Omega} \mathbb{H}^r p^r : \tau^r dx,$$

$$\langle l(t), z \rangle = \int_{\Omega} f(t) \cdot v dx + \int_{\Gamma_N} g(t) \cdot v dx,$$

$$j(z) = \int_{\Omega} \sum_{r=1}^M D^r(\tau^r) dx, \quad D^r(x) = \begin{cases} \sigma_y^r \|x\| & \text{if } \text{tr } x = 0, \\ +\infty & \text{otherwise.} \end{cases} \quad (\text{von Mises criterion})$$

Existence and uniqueness

Assumption: positive definite elastic and hardening operators:

$$\mathbb{C}\xi : \xi \geq c\|\xi\|^2 \quad \forall \xi \in \mathbb{R}^d$$

$$\mathbb{H}^r \xi : \xi \geq h^r \|\xi\|^2 \quad \forall \xi \in \mathbb{R}^d, r = 1, \dots, M$$

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Results: 1) $a(\cdot, \cdot)$ - bounded, elliptic bilinear form in \mathcal{H} ,

$$|a(w, z)| \leq \left((M + 1)\|\mathbb{C}\| + \max_{r=1, \dots, M} \|\mathbb{H}^r\| \right) \|w\|_{\mathcal{H}} \|z\|_{\mathcal{H}},$$

$$a(w, w) \geq \left(k \min_{r=1, \dots, M} \{c, h^r\} \min\{1, K\} \right) \|w\|_{\mathcal{H}}^2,$$

where $K > 0$ (Korn's first inequality) and $k = k(M) = 1 + \frac{M}{2} - \frac{1}{2}\sqrt{M(M+4)}$.

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2) $j(\cdot)$ - nonnegative, positive homogeneous and Lipschitz continuous functional in \mathcal{H}

$$|j(z_1) - j(z_2)| \leq \left(\max_{r=1, \dots, M} \{\sigma_y^r\} \text{meas}(\Omega)^{\frac{1}{2}} M^{\frac{1}{2}} \right) \|z_1 - z_2\|_{\mathcal{H}}.$$

Existence and uniqueness

Theorem: Let $l \in H^1(0, T; \mathcal{H}')$ with $l(0) = 0$. $\exists!$ $w = (u, p^1, \dots, p^M)(t) \in H^1(0, T; \mathcal{H})$ of **Problem** (Prandtl-Ishlinskii).

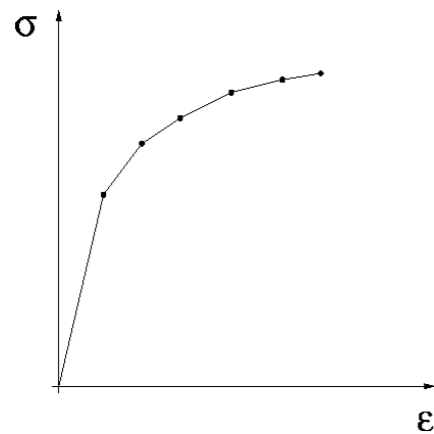
Proof: [Han, Reddy '99.]

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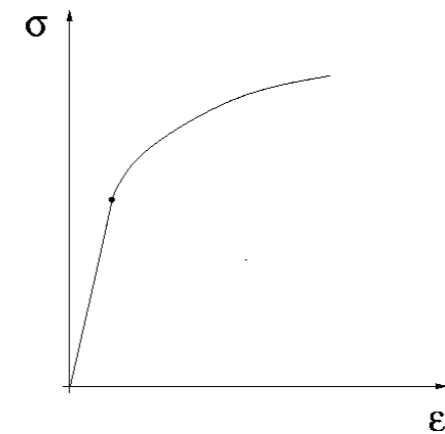
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Generalization to Measure case:



$$p = \sum_{r=1}^M p^r \quad \Rightarrow \quad p = \int_{r \in I} p(r) d\mu(r)$$

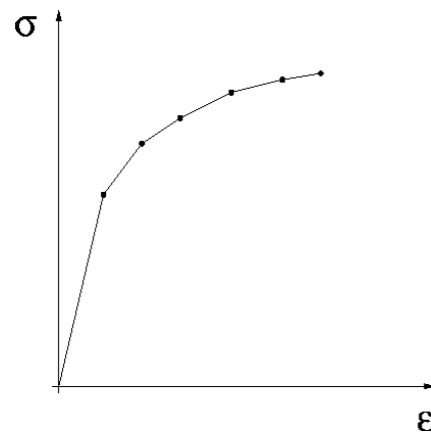


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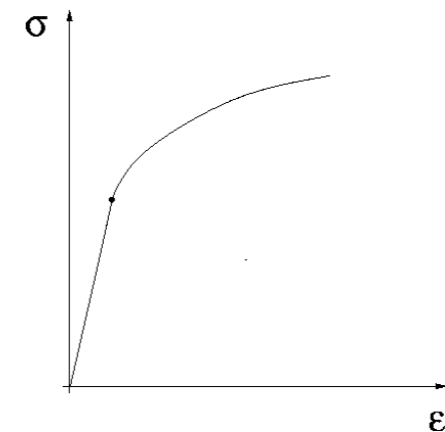
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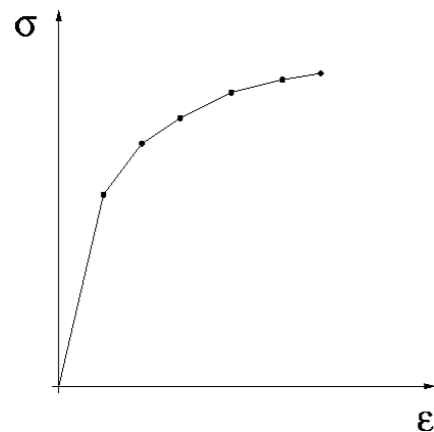
- *The same existence and uniqueness result with modified constants.*

Existence and uniqueness

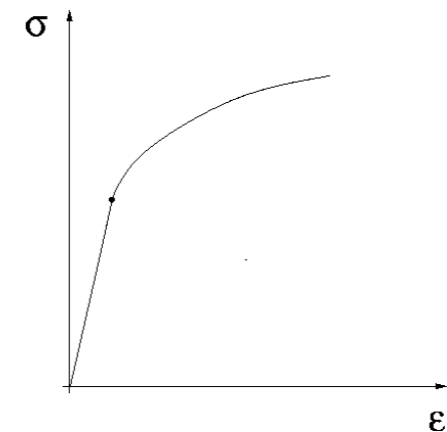
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Paper: Brokate, Carstensen, Valdman: Mathematical Modeling of the multi-yield elastoplastic material: Part 1 - Analysis

Discretization

- *in time: net* $0 = t_0 < t_1 < \dots < t_N = T$, and *implicit Euler scheme*

$$\dot{X}(t_j) = \frac{X(t_j) - X(t_{j-1})}{t_j - t_{j-1}} \quad \text{for } j = 1, \dots, N.$$

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- in space: regular triangulation \mathcal{T} of Ω
 - displacement $u : H_D^1(\Omega)$ approximated by $\mathcal{S}_D^1(\mathcal{T}) := \{v \in H_D^1(\Omega) : \forall T \in \mathcal{T}, v|_T \in \mathbb{P}_1(T)^d\}$
 - plastic strains $p_1, \dots, p_M : L^2(\Omega)$ approximated by $\mathcal{S}^0(\mathcal{T}) := \{a \in L^2(\Omega) : \forall T \in \mathcal{T}, a|_T \in \mathbb{R}\}$

Minimization problem in u, p^1, \dots, p^M

$$\begin{aligned} f(u, p^1, \dots, p^M) := & \frac{1}{2} \int_{\Omega} \mathbb{C}(\varepsilon(u) - \sum_{r=1}^M p^r) : (\varepsilon(u) - \sum_{r=1}^M p^r) dx + \frac{1}{2} \int_{\Omega} \sum_{r=1}^M |\alpha_0^r|^2 dx + \frac{1}{2} \int_{\Omega} \sum_{r=1}^M |p^r - p_0^r|^2 dx \\ & + \int_{\Omega} \sum_{r=1}^M \alpha_0^r : (p^r - p_0^r) dx + \int_{\Omega} \sum_{r=1}^M \sigma_y^r |p^r - p_0^r| dx - \int_{\Omega} bu dx \rightarrow \min. \end{aligned}$$

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New variables: $\tilde{p}^1 = p^1 - p_0^1, \dots$

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A differentiable approximation of $|\tilde{p}|$:

$$|\tilde{p}|_{\epsilon} := \begin{cases} |\tilde{p}| & \text{if } |\tilde{p}| \geq \epsilon \\ \frac{1}{2\epsilon} |\tilde{p}|^2 + \frac{\epsilon}{2} & \text{if } |\tilde{p}| < \epsilon \end{cases}$$

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Minimization strategy in each time step:

$$u^{k+1} = \operatorname{argmin}_v \min_q \bar{f}(v, q) = \operatorname{argmin}_v \tilde{f}(v, q_{opt}(v))$$

Then $p^1 = p_0^1 + \tilde{p}^1, \dots$

Direct minimization problem in u : Two-yield model

Matrix form:

$$\begin{aligned}
 f(u, \tilde{p}^1, \tilde{p}^2) = & \frac{1}{2} \begin{pmatrix} u \\ \tilde{p}^1 \\ \tilde{p}^2 \end{pmatrix}^T \begin{pmatrix} B^T \mathbb{C} B & -B^T \mathbb{C} & -B^T \mathbb{C} \\ -\mathbb{C} B & \mathbb{C} + \mathcal{D}^1 & \mathbb{C} \\ -\mathbb{C} B & \mathbb{C} & \mathbb{C} + \mathcal{D}^2 \end{pmatrix} \begin{pmatrix} u \\ \tilde{p}^1 \\ \tilde{p}^2 \end{pmatrix} + \begin{pmatrix} -b - B^T \mathbb{C} (p_0^1 + p_0^2) \\ \mathbb{C} (p_0^1 + p_0^2) + \mathbb{Q} \alpha_0^1 \\ \mathbb{C} (p_0^1 + p_0^2) + \mathbb{Q} \alpha_0^2 \end{pmatrix}^T \begin{pmatrix} u \\ \tilde{p}^1 \\ \tilde{p}^2 \end{pmatrix} \\
 & + \frac{1}{2} \mathbb{C} p_0^1 : p_0^1 + \frac{1}{2} \mathbb{C} p_0^2 : p_0^2 + \frac{1}{2} |\alpha_0^1|^2 + \frac{1}{2} |\alpha_0^2|^2 \rightarrow \min,
 \end{aligned}$$

where $\mathcal{D}^1 = \mathbb{Q} \left(1 + \frac{2\sigma_y^1}{|\tilde{p}^1|_\epsilon}\right)$, $\mathcal{D}^2 = \mathbb{Q} \left(1 + \frac{2\sigma_y^2}{|\tilde{p}^2|_\epsilon}\right)$.

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where $\mathcal{D}^1 = Q(1 + \frac{2\sigma_y^1}{|\tilde{p}^1|_\epsilon})$, $\mathcal{D}^2 = Q(1 + \frac{2\sigma_y^2}{|\tilde{p}^2|_\epsilon})$.

The Schur-Complement system in u

$$B^T (\mathbb{C} - \begin{pmatrix} \mathbb{C} \\ \mathbb{C} \end{pmatrix}^T \begin{pmatrix} \mathbb{C} + \mathcal{D}^1 & \mathbb{C} \\ \mathbb{C} & \mathbb{C} + \mathcal{D}^2 \end{pmatrix}^{-1} \begin{pmatrix} \mathbb{C} \\ \mathbb{C} \end{pmatrix}) B u = b + B^T \mathbb{C} (p_0 - \begin{pmatrix} I \\ I \end{pmatrix}^T \begin{pmatrix} \mathbb{C} + \mathcal{D}^1 & \mathbb{C} \\ \mathbb{C} & \mathbb{C} + \mathcal{D}^2 \end{pmatrix}^{-1} \begin{pmatrix} \mathbb{C} p_0 + Q\alpha_0^1 \\ \mathbb{C} p_0 + Q\alpha_0^2 \end{pmatrix})$$

where $p_0 = p_0^1 + p_0^2$. \Rightarrow **Multigrid-PCG**

Direct minimization problem in \tilde{p} : analytical approach

Kinematic hardening model ($M = 1$):

$$f(q) = \frac{1}{2}(\mathbb{C} + \mathbb{H})q : q - q : A + \sigma_y \|q\| \rightarrow \min$$

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Two-yield hardening model ($M = 2$):

$$f \begin{pmatrix} q^1 \\ q^2 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} \mathbb{C} + \mathbb{H}^1 & \mathbb{C} \\ \mathbb{C} & \mathbb{C} + \mathbb{H}^2 \end{pmatrix} \begin{pmatrix} q^1 \\ q^2 \end{pmatrix} : \begin{pmatrix} q^1 \\ q^2 \end{pmatrix} - \begin{pmatrix} q^1 \\ q^2 \end{pmatrix} : \begin{pmatrix} A^1 \\ A^2 \end{pmatrix} + \sigma_y^1 \|q^1\| + \sigma_y^2 \|q^2\| \rightarrow \min$$

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$$\text{minimizer } (\tilde{p}^1, \tilde{p}^2) = ?$$

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minimizer $(\tilde{p}^1, \tilde{p}^2) = ?$

$\tilde{p}^2 \neq 0 \Rightarrow \|\tilde{p}^2\|$ is a root of a **6-th** degree polynomial.

Symbolic methods (Gröbner basis?)

Direct minimization problem in \tilde{p} : Two-yield model - iterative approach

Algorithm (*): Given tolerance ≥ 0 .

(a) Choose $(p_1^0, p_2^0) \in \text{dev } \mathbb{R}_{sym}^{d \times d} \times \text{dev } \mathbb{R}_{sym}^{d \times d} \xrightarrow{\min!d}$, set $i := 0$.

(b) Find $p_2^{i+1} \in \text{dev } \mathbb{R}_{sym}^{d \times d}$ s. t.

$$f(p_1^i, p_2^{i+1}) = \min_{Q_2 \in \text{dev } \mathbb{R}_{sym}^{d \times d}} f(p_1^i, Q_2).$$

(c) Find $p_1^{i+1} \in \text{dev } \mathbb{R}_{sym}^{d \times d}$ s. t.

$$f(p_1^{i+1}, p_2^{i+1}) = \min_{q_1 \in \text{dev } \mathbb{R}_{sym}^{d \times d}} f(q_1, p_2^{i+1}).$$

(d) If $\frac{\|p_1^{i+1} - p_1^i\| + \|p_2^{i+1} - p_2^i\|}{\|p_1^{i+1}\| + \|p_1^i\| + \|p_2^{i+1}\| + \|p_2^i\|} > \text{tolerance}$ set $i := i + 1$ and goto (b), otherwise output (p_1^{i+1}, p_2^{i+1}) .

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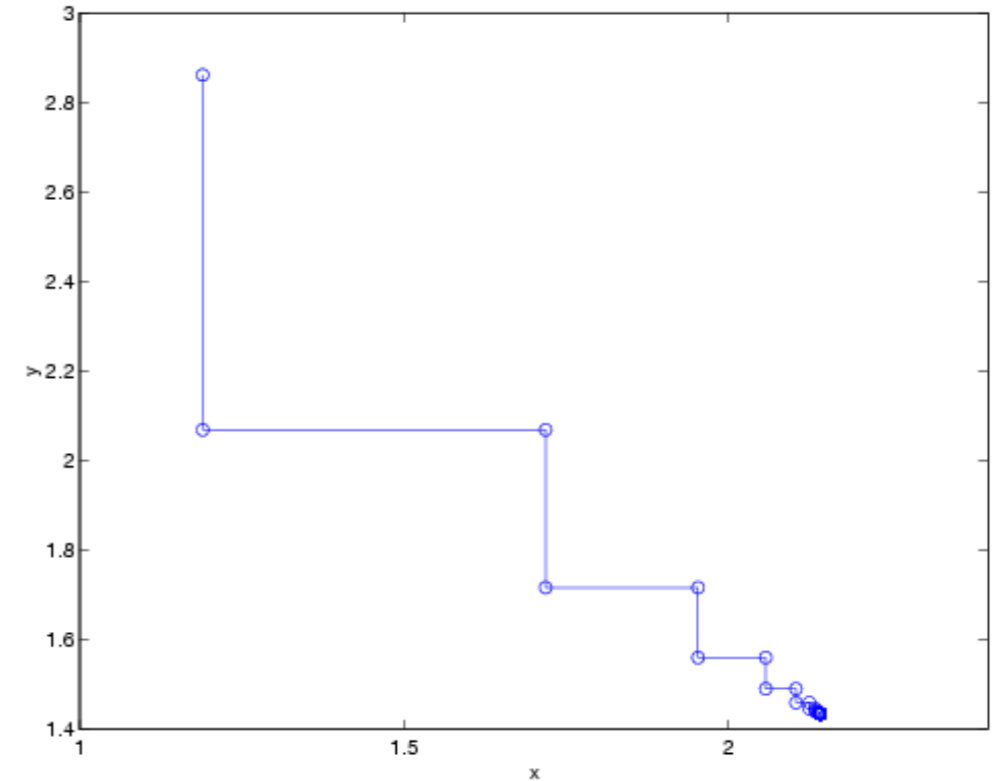
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$$f(p_1^{i+1}, p_2^{i+1}) = \min_{q_1 \in \text{dev } \mathbb{R}_{sym}^{d \times d}} f(q_1, p_2^{i+1}).$$

(d) If $\frac{\|p_1^{i+1} - p_1^i\| + \|p_2^{i+1} - p_2^i\|}{\|p_1^{i+1}\| + \|p_1^i\| + \|p_2^{i+1}\| + \|p_2^i\|} > \text{tolerance}$ set $i := i + 1$ and goto (b), otherwise output (p_1^{i+1}, p_2^{i+1}) .

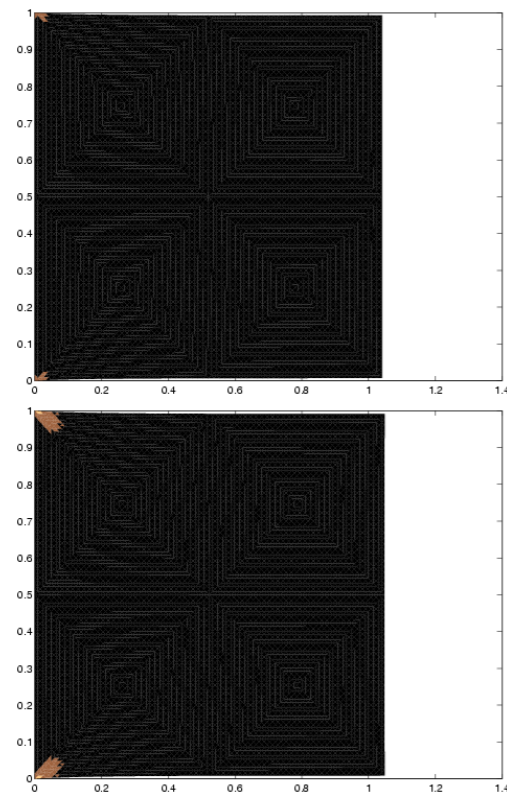


The approximations $p_1^i = (x^i, 0; 0, -x^i)$, $p_2^i = (y^i, 0; 0, -y^i)$, $i = 0, 1, \dots$ of Algorithm (*) in the $x - y$ coordinate system.

- global convergence with the rate 1/2: $\|p_1^i - p_1\|^2 + \|p_2^i - p_2\|^2 \leq C_0 \cdot q^i$

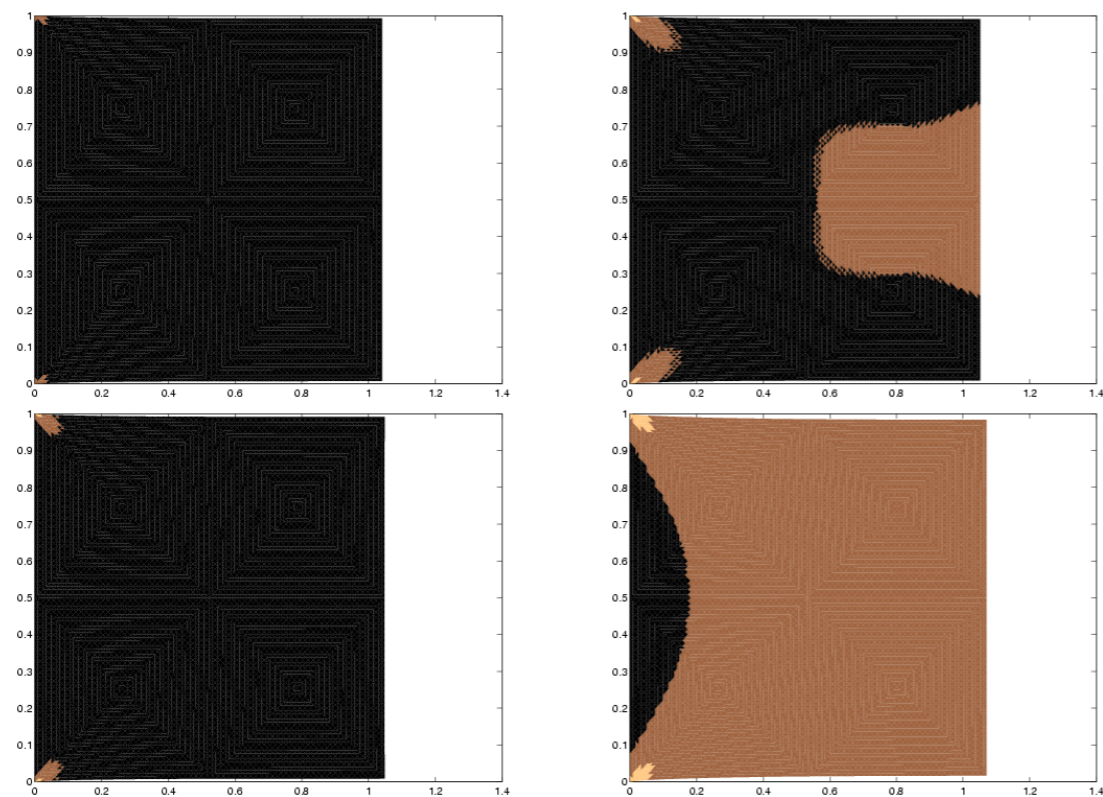
Matlab calculations in 2D

Elastoplastic domains (black - elastic, brown - first plastic, yellow - second plastic)



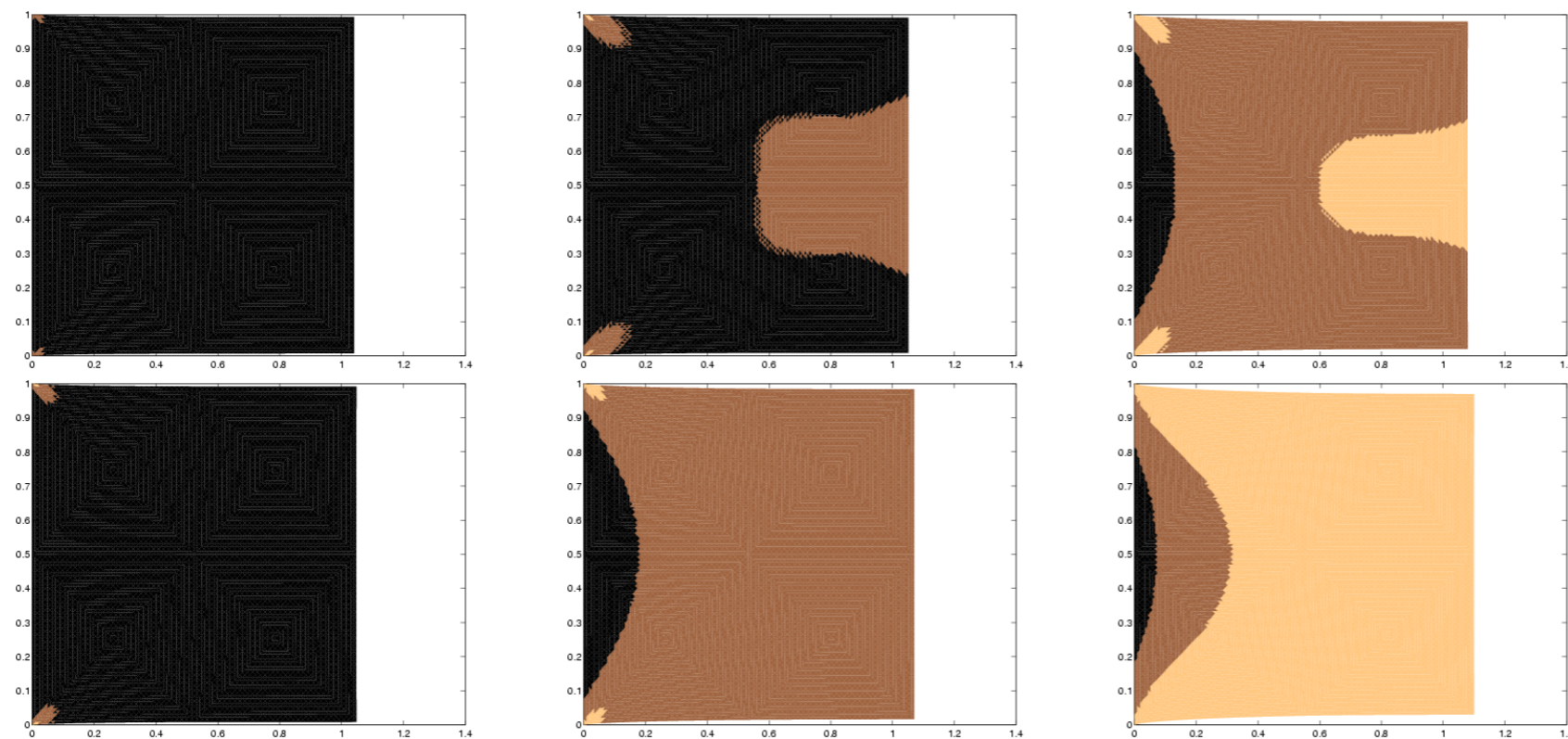
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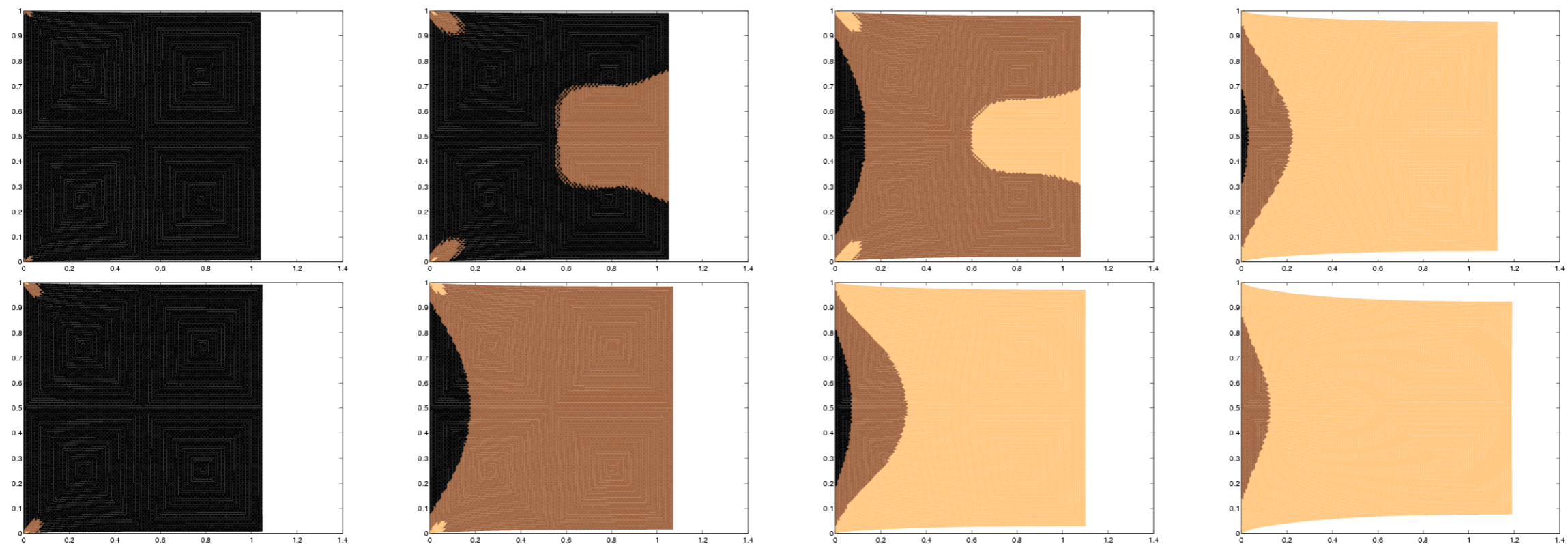
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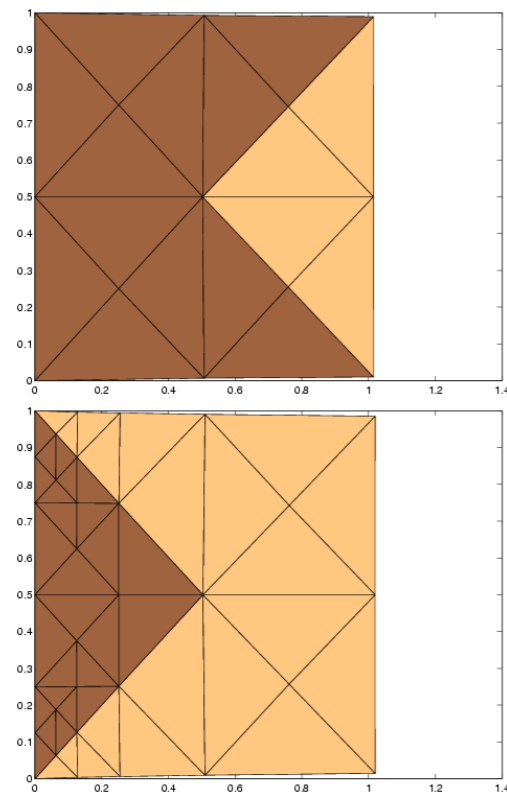
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Evolution of elastoplastic zones at 8 graduate discrete times of the two-yield beam. Calculated for 16334 elements, CPU time = 25.17 hours.

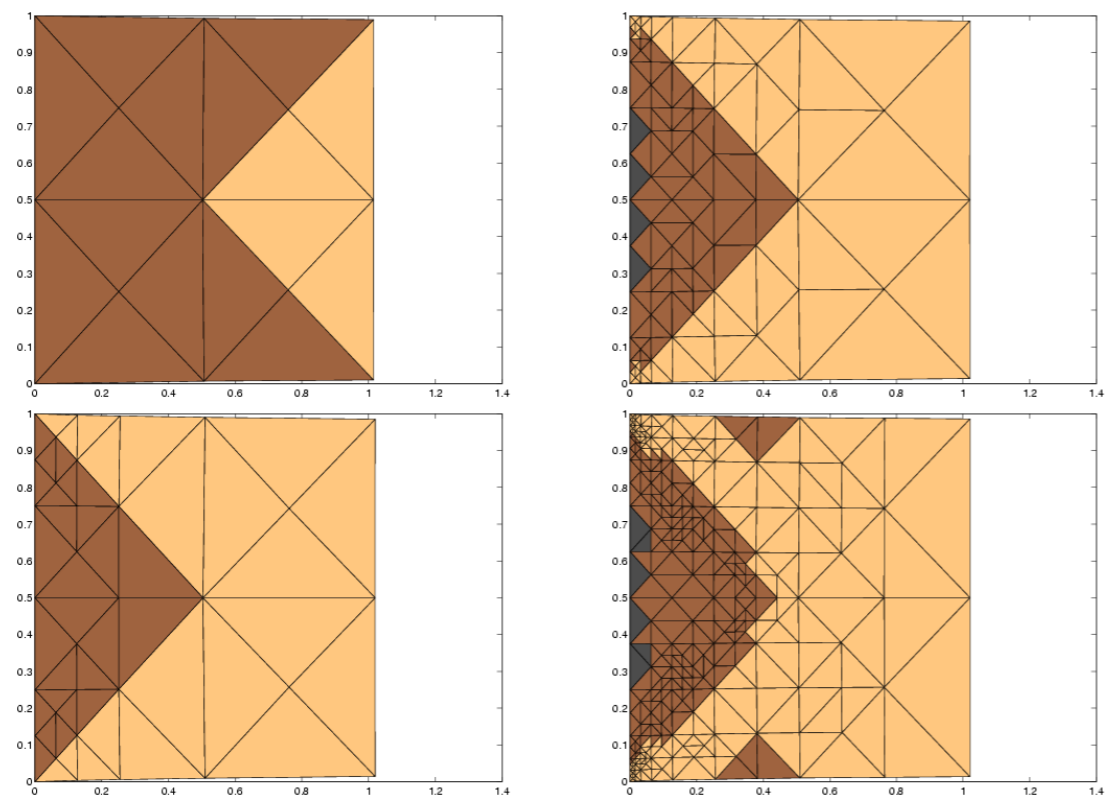
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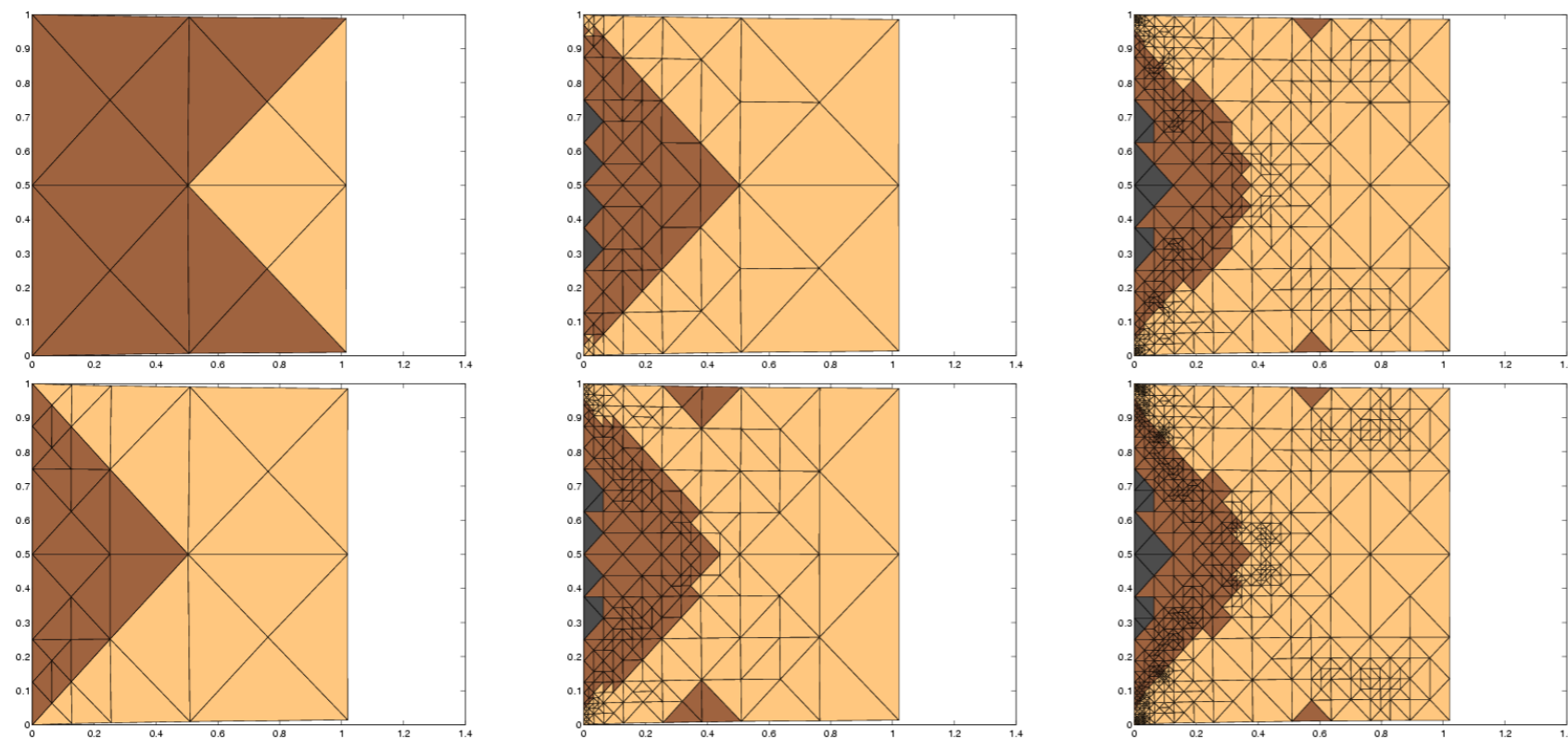
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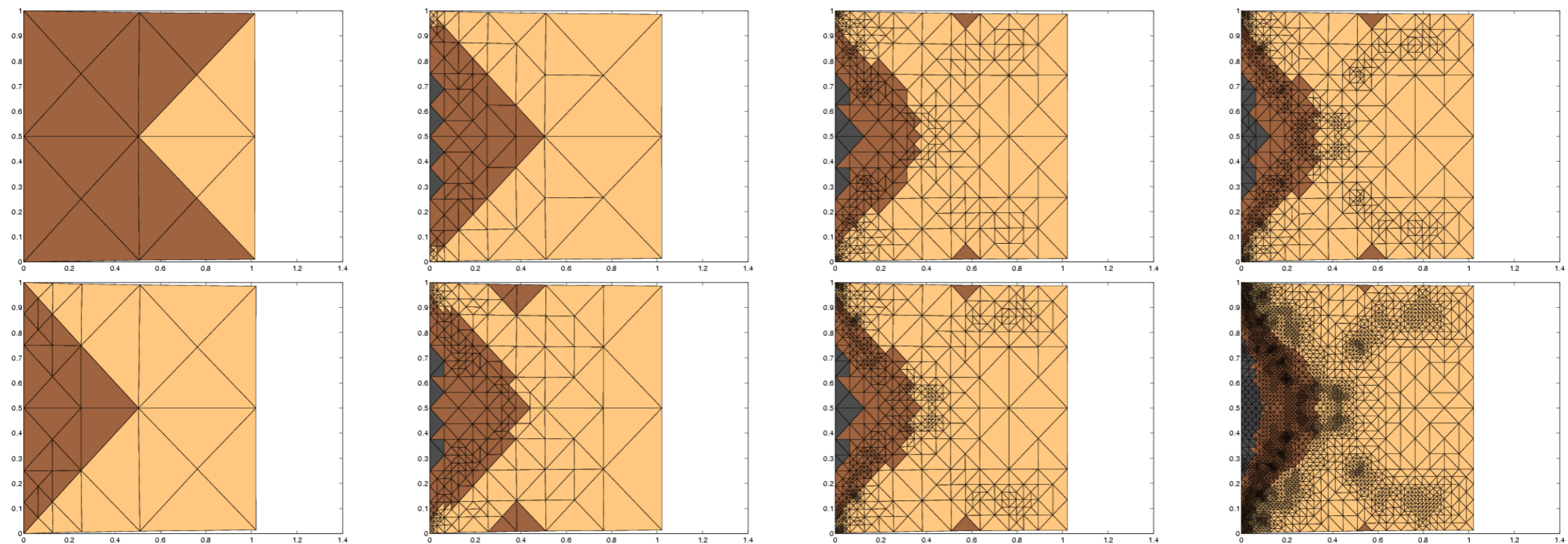
Matlab calculations in 2D

Elastoplastic domains (black - elastic, brown - first plastic, yellow - second plastic)



Matlab calculations in 2D

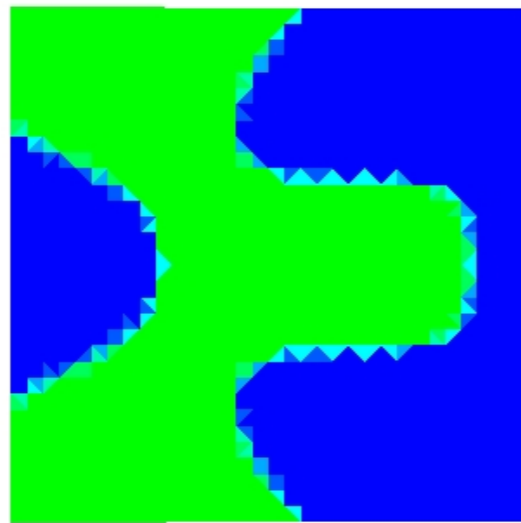
Elastoplastic domains (black - elastic, brown - first plastic, yellow - second plastic)



ZZ- adaptively refined meshes and elastoplastic zones, one time-step problem, two-yield beam.

NGSOLVE calculations in 2D

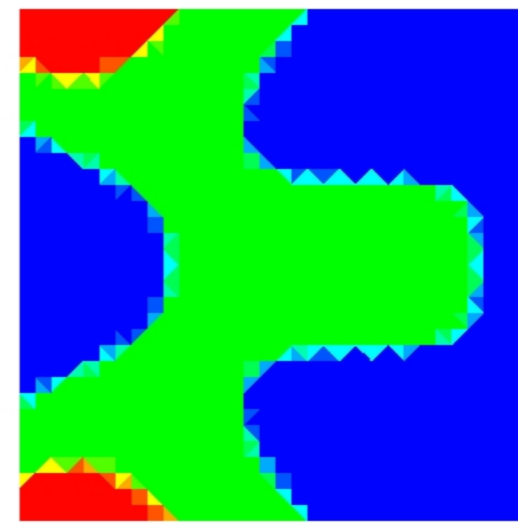
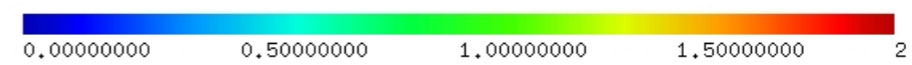
Elastoplastic domains (blue - elastic, green - first plastic, red - second plastic)



y
x

Netgen 4.2

Kinematic hardening model.



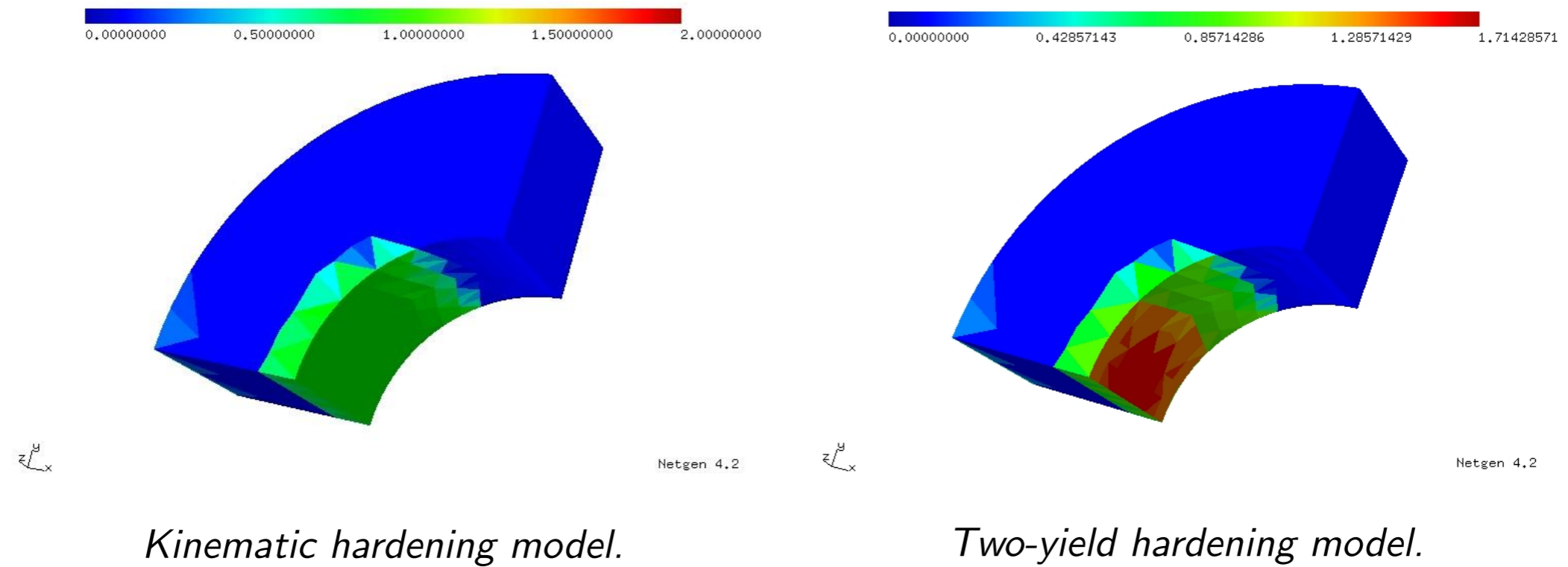
y
x

Netgen 4.2

Two-yield hardening model.

NGSOLVE calculations in 3D

Elastoplastic domains (blue - elastic, green - first plastic, red - second plastic)



Outlook

Future Work:

- *Convergence proof*

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Download Netgen and NGSolve:

<http://www.hpfem.jku.at/netgen/>