

# MIXED FINITE ELEMENT METHOD FOR A DEGENERATE CONVEX VARIATIONAL PROBLEM FROM TOPOLOGY OPTIMISATION

CARSTEN CARSTENSEN <sup>†¶</sup>, DAVID GÜNTHER <sup>‡</sup>, AND HELLA RABUS<sup>§</sup> <sup>||</sup>

**Key words.** AFEM, adaptive mixed finite element method, Optimal design, degenerate convex minimisation

**Abstract.** The optimal design task of this paper seeks the distribution of two materials of prescribed amounts for maximal torsion stiffness of an infinite bar of given cross section. This example of relaxation in topology optimisation leads to a degenerate convex minimisation problem

$$E(v) := \int_{\Omega} \varphi_0(|\nabla v|) dx - \int_{\Omega} f v dx \quad \text{for } v \in V := H_0^1(\Omega)$$

with possibly multiple primal solutions  $u$ , but with unique stress

$$\sigma := \varphi_0'(|\nabla u|) \text{sign } \nabla u.$$

The mixed finite element method is motivated by the smoothness of the stress variable  $\sigma \in H_{\text{loc}}^1(\Omega; \mathbb{R}^2)$  while the primal variables are un-controllable and possibly non-unique. The corresponding nonlinear mixed finite element method is introduced, analysed, and implemented.

The striking result of this paper is a sharp a posteriori error estimation in the dual formulation, while the a posteriori error analysis in the primal problem suffers from the reliability-efficiency gap. An empirical comparison of that primal with the new mixed discretisation schemes is intended for uniform and adaptive mesh-refinements.

**1. Introduction.** This paper appears to be the first attempt to utilise mixed finite element methods (MFEMs) for degenerate minimisation problems in the calculus of variations. The usage of MFEM in relaxed formulations for macroscopic simulations in computational microstructures [3, 7, 9, 22] is motivated by the properties of the primal and dual variables. The primal variables (e.g., a deformation or displacement) may be non-unique [17] or less regular, while the dual (e.g., a flux or stress) variable is unique and locally smooth [6]. Hence a mixed scheme, which relies on smooth dual variables, might enjoy superior convergence properties.

The model problem is motivated by an optimal design problem, where a given domain  $\Omega \subset \mathbb{R}^2$  has to be filled with two materials of different elastic shear stiffnesses with energy

$$E(v) := \int_{\Omega} \varphi_0(|\nabla v|) dx - \int_{\Omega} f v dx \quad \text{for } v \in V := H_0^1(\Omega) \quad (1.1)$$

for given right-hand side  $f \in L^2(\Omega)$  and energy density function  $\varphi_0 \in C^0([0, \infty); \mathbb{R})$  of Section 2.

The model has been analysed in [18, 11, 20, 21] and computed in [19, 6]. Recently, a convergent adaptive finite element method in its primal form has been introduced in [4].

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<sup>†</sup>Humboldt-Universität zu Berlin, 10099 Berlin, Germany; Department of Computational Science and Engineering, Yonsei University, 120-749 Seoul, Korea; (cc@mathematik.hu-berlin.de)

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<sup>‡</sup>Max-Planck-Institut für Informatik, 66123 Saarbrücken, Germany; (david.guenther@zib.de)

<sup>§</sup>Humboldt-Universität zu Berlin, 10099 Berlin; Germany; (rabus@mathematik.hu-berlin.de);

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While the solutions of the primal and dual problem coincide in the continuous case, this does not need to be true for discrete calculations in general. In the dual formulation, we avoid the difficulties arising from the fact, that the gradient of the energy density functional  $\varphi'_0$  is not strongly monotone. This may lead to multiple primal variables  $u$ , while there is a unique stress-type variable  $\sigma := \varphi'_0(|\nabla u|) \operatorname{sign} \nabla u$  [6, 8, 4]. In contrast to the continuous differentiability of  $\varphi'_0$ , its conjugate function  $\varphi_0^*$  is solely Lipschitz-continuous. To overcome the lack of differentiability we approximate  $\varphi_0^*$  by its Yosida regularisation  $\varphi_\varepsilon^*$ .

The proposed mixed formulation is based on the dual formulation: Seek  $(u, \sigma) \in L^2(\Omega) \times H(\operatorname{div}, \Omega)$  with

$$\operatorname{div} \sigma + f = 0 \text{ and } \nabla u \in \partial\Phi_0^*(\sigma) \text{ in } \Omega. \quad (\mathbf{D})$$

The discretisation is based on piecewise polynomial subspaces  $RT_0(\mathcal{T}) \subseteq H(\operatorname{div}, \Omega)$  and  $\mathcal{P}_0(\mathcal{T}) \subseteq L^2(\Omega)$  named after Raviart and Thomas and introduced in Section 3. For  $\varepsilon > 0$  piecewise constant with respect to  $\mathcal{T}$ , the discrete regularised dual problem reads: Seek  $(u_{\varepsilon h}, \sigma_{\varepsilon h}) \in \mathcal{P}_k(\mathcal{T}) \times RT_k(\mathcal{T})$ , such that for all  $(v_h, \tau_h) \in \mathcal{P}_k(\mathcal{T}) \times RT_k(\mathcal{T})$ , it holds that

$$\begin{aligned} (\tau_h, D\Phi_\varepsilon^*(\sigma_{\varepsilon h}))_{L^2(\Omega)} + (u_{\varepsilon h}, \operatorname{div} \tau_h)_{L^2(\Omega)} &= 0, \\ (v_h, \operatorname{div} \sigma_{\varepsilon h})_{L^2(\Omega)} + (f, v_h)_{L^2(\Omega)} &= 0. \end{aligned} \quad (\mathbf{D}_{\varepsilon h})$$

The main theorems in Section 3 verify that poor a priori error estimates are caused by the lack of smoothness, while efficient and reliable a posteriori error estimates are derived. Numerical simulations show that the convergence of the adaptive scheme is improved in the presence of geometric singularities such as nonconvex corners. Furthermore, compared to the primal formulation as considered in [4], the experiments of Section 6 of the regularised dual mixed form reveal reduced convergence rates but no efficiency-reliability gap.

The remaining parts of the paper are organised as follows. Section 2 covers a preliminary analysis of the model problem and its energy density function. The regularised and discrete mixed formulation of the problem is introduced in Section 3, followed by the investigation of the existence and uniqueness of the exact and discrete solutions in Section 4. Section 5 presents a throughout a priori and a posteriori error analysis. The adaptive mesh-refining algorithm and some numerical experiments conclude the paper in Section 6.

In this paper we follow the standard notation for the Lebesgue  $L^2(\Omega)$ ,  $L^2(\Omega; \mathbb{R}^2)$  and Sobolev spaces  $H^1(\Omega)$ ,  $H^1(\Omega; \mathbb{R}^2)$ ;  $H(\operatorname{div}, \Omega)$  denotes the Hilbert space of  $L^2$ -functions with square-integrable divergence. The  $L^2(\Omega)$  scalar product is abbreviated by  $(\cdot, \cdot)_{L^2(\Omega)}$ , while  $\langle \cdot, \cdot \rangle$  denotes the scalar product in  $\mathbb{R}^n$ .

## 2. Preliminaries.

**2.1. An optimal design problem.** The task is to seek the distribution of two materials of fixed volume fraction in the cross section of an infinite bar given by the domain  $\Omega \subseteq \mathbb{R}^2$  for maximal torsion stiffness. The focus of this paper lies on the analysis and numerical studies of the variational problem, while the precise mathematical modelling may divert from the emphasis of this paper. For details on the mathematical modelling the reader is referred to [4, Section 2] and the references given in Section 1.

Let  $0 < t_1 < t_2$  and the reciprocal shear stiffness  $0 < \mu_1 < \mu_2 < \infty$  with  $t_1\mu_2 = \mu_1 t_2$ , and  $0 < \xi < 1$  representing the ratio of amounts of the two materials,

$|\Omega_1| = \xi |\Omega|$ ,  $|\Omega_2| = \Omega - |\Omega_1|$  and  $t_1 = \sqrt{2\lambda\mu_1/\mu_2}$ . The Lagrange parameter  $\lambda \in \mathbb{R}$  is fixed for a specific geometry  $\Omega$  and the choice of  $\xi$  [4, 14, 17, 18, 19].

In the relaxed formulation of the model from [19], the right-hand side  $f \equiv 1$  is constant and the locally Lipschitz continuous energy density function  $\varphi_0 : [0, \infty) \rightarrow \mathbb{R}$  reads

$$\varphi_0(t) = \lambda\xi(\mu_1 - \mu_2) + \begin{cases} \frac{\mu_2}{2}t^2 & \text{for } 0 \leq t \leq t_1, \\ t_1\mu_2(t - \frac{t_1}{2}) & \text{for } t_1 \leq t \leq t_2, \\ \frac{\mu_1}{2}t^2 + \frac{\mu_1 t_2}{2}(t_2 - t_1) & \text{for } t_2 \leq t. \end{cases}$$

Thus, the primal formulation is the minimisation of  $E$  in (1.1). There exists minimisers of  $E$  which are not necessarily unique. For  $f \in L^2(\Omega)$ , the stress field  $\sigma := \varphi_0'(|\nabla u|) \text{sign } \nabla u$  is unique and locally smooth, i.e.,  $\sigma \in H_{\text{loc}}^1(\Omega; \mathbb{R}^2)$  while  $f \in H_0^1(\Omega)$  (excluded in this work) implied  $\sigma \in H^1(\Omega, \mathbb{R}^2)$ , cf. [6].

**2.2. Dual functional and Yosida regularisation.** Direct calculations lead to the dual function  $\varphi_0^*$  of  $\varphi_0$  and its Yosida regularisation  $\varphi_\varepsilon^*$  as stated in the following Lemma. We use standard notation of convex analysis [23].

LEMMA 2.1. *The dual (or conjugate) function  $\varphi_0^*$  of  $\varphi_0$  reads*

$$\varphi_0^*(t) = -\lambda\xi(\mu_1 - \mu_2) + \begin{cases} \frac{t^2}{2\mu_2} & \text{for } t \leq t_1\mu_2, \\ \frac{t^2}{2\mu_1} - \frac{\mu_1 t_2^2}{2} + \frac{t_1^2 \mu_2}{2} & \text{for } t_1\mu_2 \leq t. \end{cases}$$

*It is piecewise polynomial and globally convex, and Lipschitz continuous on compact subsets but not differentiable at  $t = t_1\mu_2 = \mu_1 t_2$ .*

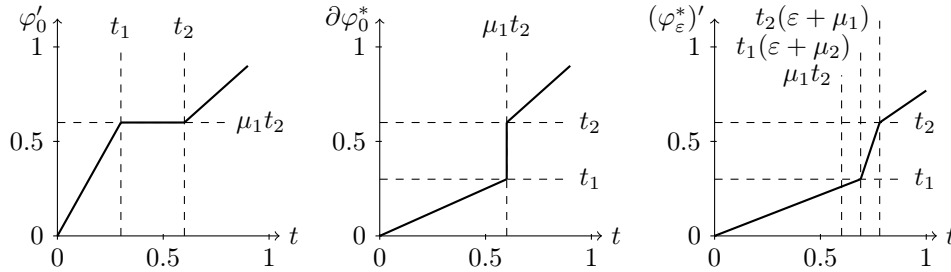
*For fixed  $\varepsilon > 0$  and all  $t \geq 0$ , the Yosida regularisation  $\varphi_\varepsilon^*$  of  $\varphi_0^*$  is defined by  $\varphi_\varepsilon^*(t) := \inf_{z \in \mathbb{R}} \left( \varphi_0^*(z) + \frac{1}{2\varepsilon} |t - z|^2 \right)$  and equals*

$$\varphi_\varepsilon^*(t) = -\lambda\xi(\mu_1 - \mu_2) + \begin{cases} \frac{t^2}{2(\varepsilon + \mu_2)} & \text{for } 0 \leq t < t_1(\varepsilon + \mu_2), \\ \frac{\mu_2}{2}t_1^2 + \frac{1}{2\varepsilon} |t_1\mu_2 - t|^2 & \text{for } t_1\mu_2 + \varepsilon t_1 \leq t \leq t_1\mu_2 + \varepsilon t_2, \\ \frac{t^2}{2(\mu_1 + \varepsilon)} - \frac{\mu_1 t_2^2}{2} + \frac{t_1^2 \mu_2}{2} & \text{for } t_2(\mu_1 + \varepsilon) < t. \end{cases}$$

*Let  $C_\mu := \frac{1}{\mu_1^2} + \frac{1}{2\mu_2^2}$ . Then, the difference of  $\varphi_0^*$  and  $\varphi_\varepsilon^*$  is bounded in the sense that*

$$0 \leq \sup_{z \in \mathbb{R}} (\varphi_0^*(t) - \varphi_0^*(z) - \frac{1}{2\varepsilon} |t - z|^2) = \varphi_0^*(t) - \varphi_\varepsilon^*(t) \leq C_\mu \varepsilon t^2 \leq \mathcal{O}(\varepsilon) t^2. \quad \square$$

The function  $\varphi_\varepsilon^*$  is differentiable, hence the subgradient  $\partial\varphi_\varepsilon^*(a) = \{(\varphi_\varepsilon^*)'(a)\}$  is a singleton, while  $\varphi_0^*$  is not smooth and  $\partial\varphi_0^*(t_1\mu_2) = [t_1, t_2]$  is a compact interval. The differentials  $\varphi_0'$ ,  $\partial\varphi_\varepsilon^*$ , and  $(\varphi_\varepsilon^*)'$  are depicted in the following sketch.



**2.3. Remarks on  $\varphi_\varepsilon$ ,  $\varphi_\varepsilon^*$  and  $\Phi_\varepsilon$ ,  $\Phi_\varepsilon^*$ .** The energy density function

$$\Phi_\varepsilon : \mathbb{R}^n \rightarrow \mathbb{R}, \quad \Phi_\varepsilon(F) := \varphi_\varepsilon(|F|) \text{ for all } F \in \mathbb{R}^n,$$

its dual and regularised dual function enjoy the following properties.

**(i)  $\Phi_\varepsilon$  and  $\Phi_\varepsilon^*$ .** For any  $\varepsilon > 0$  the function  $\Phi_\varepsilon := \varphi_\varepsilon(|\cdot|)$ , satisfies

$$D\Phi_\varepsilon(F) = \varphi'_\varepsilon(|F|) \text{ sign } F \text{ for all } F \in \mathbb{R}^n$$

with the unit ball  $B(0, 1) := \{x \in \mathbb{R}^n \mid |x| \leq 1\}$  and

$$\text{sign } F := \begin{cases} B(0, 1) & \text{if } |F| = 0, \\ F/|F| & \text{otherwise.} \end{cases}$$

Notice that  $\varphi'_\varepsilon(0) = 0$  and  $\varphi_\varepsilon(0) = \varphi_0(0) = 0$  imply

$$D\Phi_\varepsilon(F) = \begin{cases} \varphi'_\varepsilon(|F|)F/|F| & \text{if } |F| \neq 0, \\ 0 & \text{otherwise.} \end{cases}$$

For  $\varepsilon > 0$  and all  $F \in \mathbb{R}^n$ , the dual of  $\Phi_\varepsilon = \varphi_\varepsilon(|\cdot|)$  reads

$$\begin{aligned} \Phi_\varepsilon^*(F) &= \varphi_\varepsilon^*(|F|) \text{ for all } F \in \mathbb{R}^n \text{ and} \\ D\Phi_\varepsilon^*(F) &= (\varphi_\varepsilon^*)'(|F|) \text{ sign } F. \end{aligned}$$

For  $\varepsilon = 0$ ,  $\Phi_0^* := \varphi_0^*(|\cdot|)$  satisfies

$$\partial\Phi_0^*(F) = \partial\varphi_0^*(|F|) \text{ sign } F \text{ for all } F \in \mathbb{R}^n.$$

**(ii) Convexity control for  $\Phi_0$ .** The function  $\Phi_0$  allows convexity control in the sense that for all  $a, b \in \mathbb{R}^n$ ,  $A \in \partial\Phi_0(a)$ , and for all  $B \in \partial\Phi_0(b)$ , it holds that

$$\frac{1}{\mu_2} |A - B|^2 \leq \langle A - B, a - b \rangle.$$

**(iii) Strong monotonicity of  $\partial\Phi_\varepsilon^*$ .** The subgradient  $\partial\Phi_\varepsilon^*$  is strongly monotone in the sense that, with  $C_M := \mu_2 + \varepsilon$  and  $\varepsilon \geq 0$ , it holds that

$$\mu_2 |a - b|^2 \leq C_M |a - b|^2 \leq \langle \partial\Phi_\varepsilon^*(a) - \partial\Phi_\varepsilon^*(b), a - b \rangle \text{ for all } a, b \in \mathbb{R}^n.$$

**(iv) Strong convexity of  $\Phi_\varepsilon^*$ .** For all  $\varepsilon \geq 0$  the strong monotonicity of  $\partial\Phi_\varepsilon^*$  and the definition of the subdifferential lead to

$$2\mu_2 |a - b|^2 \leq 2C_M |a - b|^2 \leq \langle \partial\Phi_\varepsilon^*(a), a - b \rangle - \Phi_\varepsilon^*(a) + \Phi_\varepsilon^*(b)$$

for all  $a, b \in \mathbb{R}^n$ , cf. [16, Thm. D2.6.1]. Hence,  $\Phi_\varepsilon^*$  is strongly convex.

**(v) Lipschitz continuity of  $(\varphi_\varepsilon^*)'$ .** For all  $\varepsilon > 0$ ,  $\varphi_\varepsilon^*$  is continuously differentiable and

$$(\varphi_\varepsilon^*)'(t) = \begin{cases} \frac{t}{\mu_2 + \varepsilon} & \text{for } t < t_1(\mu_2 + \varepsilon), \\ \frac{t - t_1\mu_2}{\varepsilon} & \text{for } t_1\mu_2 + \varepsilon t_1 \leq t \leq t_1\mu_2 + \varepsilon t_2, \\ \frac{t}{\mu_1 + \varepsilon} & \text{for } t_2(\mu_1 + \varepsilon) < t \end{cases}$$

is Lipschitz continuous with Lipschitz constant  $\text{Lip}(D\varphi_\varepsilon^*) = 1/\varepsilon$ .

**(vi) Discontinuity of  $\partial\Phi_0^*$ .** The subgradient  $\partial\Phi_0^*$  is piecewise Lipschitz continuous and jumps at  $|z| = \mu_1 t_2$ . However, the following estimate holds

$$|\partial\Phi_0^*(a) - \partial\Phi_0^*(b)| \leq \delta(a, b) + |a - b|/\mu_1$$

for all  $a, b \in \mathbb{R}^n$  with

$$\delta(a, b) := \begin{cases} t_2 - t_1 & \text{if } \min\{|a|, |b|\} \leq t_1 \mu_2 \leq \max\{|a|, |b|\}, \\ 0 & \text{otherwise.} \end{cases}$$

This estimate can be extended to  $\Phi_\varepsilon^*$  and  $\varepsilon \geq 0$  in the sense that

$$\begin{aligned} |\partial\Phi_\varepsilon^*(a) - \partial\Phi_\varepsilon^*(b)| &\leq \delta_\varepsilon(a, b) + |a - b|/(\mu_1 + \varepsilon) \\ &\leq \delta_\varepsilon(a, b) + |a - b|/\mu_1 \end{aligned}$$

for all  $a, b \in \mathbb{R}^n$  with

$$\delta_\varepsilon(a, b) := \begin{cases} t_2 - t_1 & \text{if } \exists t \in t_1 \mu_2 + \varepsilon[t_1, t_2] \text{ such that} \\ & \min\{|a|, |b|\} \leq t \leq \max\{|a|, |b|\}, \\ 0 & \text{otherwise.} \end{cases}$$

### 3. Mixed formulation and its discretisations.

**3.1. Motivation for mixed formulation.** The direct method of calculus of variations yields the existence of a minimiser  $u$  of  $E$  in  $V := H_0^1(\Omega)$  with

$$E(v) := \int_{\Omega} (\varphi_0(|\nabla v|) - fv) \, dx. \quad (3.1)$$

The exact stress  $\sigma = \varphi_0'(|\nabla u|) \text{sign}(\nabla u)$  satisfies the equilibrium  $\text{div } \sigma + f = 0$  in  $\Omega$  as the strong form of the Euler-Lagrange equations. Given any right-hand side  $f \in L^2(\Omega)$  and the convex  $C^1$ -functional  $\Phi_0 : \mathbb{R}^n \rightarrow \mathbb{R}$ , the pair  $(u, \sigma) \in V \times H(\text{div}, \Omega)$  solves the primal mixed formulation

$$\text{div } \sigma + f = 0 \text{ and } \sigma = \text{D}\Phi_0(\nabla u) \text{ in } \Omega. \quad (\mathbf{P})$$

By duality of convex functions it holds, for all  $\alpha, a \in \mathbb{R}^n$ , that

$$\alpha \in \partial\Phi_0(a) \Leftrightarrow a \in \partial\Phi_0^*(\alpha).$$

This allows the reformulation

$$\sigma = \text{D}\Phi_0(\nabla u) \Leftrightarrow \nabla u \in \partial\Phi_0^*(\sigma).$$

Consequently, **(P)** reads in terms of the conjugated functional as

$$\text{div } \sigma + f = 0 \text{ and } \nabla u \in \partial\Phi_0^*(\sigma) \text{ in } \Omega. \quad (\mathbf{D})$$

**3.2. Regularised mixed formulation.** For a  $C^1$ -regularisation  $\varphi_\varepsilon^*$  of  $\varphi_0^*$  with  $\varphi_\varepsilon^* \rightarrow \varphi_0^*$  as  $\varepsilon \rightarrow 0$ , the regularised problem of **(D)** reads: Given any  $\varepsilon > 0$  piecewise constant on  $\mathcal{T}$ , seek  $(u_\varepsilon, \sigma_\varepsilon) \in V \times H(\text{div}, \Omega)$  with

$$\text{div } \sigma_\varepsilon + f = 0 \text{ and } \nabla u_\varepsilon = \text{D}\Phi_\varepsilon^*(\sigma_\varepsilon) \text{ in } \Omega.$$

The corresponding weak mixed formulation **(D $_\varepsilon$ )** reads: Seek  $(u_\varepsilon, \sigma_\varepsilon) \in L^2(\Omega) \times H(\text{div}, \Omega)$  such that for all  $(v, \tau) \in L^2(\Omega) \times H(\text{div}, \Omega)$  it holds that

$$\begin{aligned} (\tau, \text{D}\Phi_\varepsilon^*(\sigma_\varepsilon))_{L^2(\Omega)} + (u_\varepsilon, \text{div } \tau)_{L^2(\Omega)} &= 0, \\ (v, \text{div } \sigma_\varepsilon)_{L^2(\Omega)} + (f, v)_{L^2(\Omega)} &= 0. \end{aligned} \quad (\mathbf{D}_\varepsilon)$$

**3.3. Discrete formulation.** Given a shape-regular triangulation  $\mathcal{T}$  into triangles  $T$  of  $\Omega$  which covers  $\bar{\Omega} = \cup_{T \in \mathcal{T}} T$  exactly, let  $\mathcal{E}$  denote the set of edges  $E$  of  $\mathcal{T}$  and  $\mathcal{E}(\Omega)$  the set of interior edges. For any  $k = 0, 1, 2, \dots$ , set

$$\begin{aligned} \mathcal{P}_k(T) &:= \{\text{polynomials on } T \text{ of total degree } \leq k\}, \\ \mathcal{P}_k(\mathcal{T}) &:= \{v \in L^2(\Omega) \mid v|_T \in \mathcal{P}_k(T) \text{ for all } T \in \mathcal{T}\}, \\ \mathcal{S}_k(\mathcal{T}) &:= \{v \in \mathcal{P}_k(\mathcal{T}) \mid v \text{ globally continuous in } \bar{\Omega}\}, \\ \mathcal{S}_{k,0}(\mathcal{T}) &:= \{v \in \mathcal{S}_k(\mathcal{T}) \mid v = 0 \text{ on } \partial\Omega\}, \\ RT_k(T) &:= \left\{ (x, y) \mapsto \begin{pmatrix} p_1(x, y) \\ p_2(x, y) \end{pmatrix} + p_3(x, y) \begin{pmatrix} x \\ y \end{pmatrix} \mid p_1, p_2, p_3 \in \mathcal{P}_k(T) \right\}. \end{aligned}$$

Let  $[p]_E := p|_{T_+} - p|_{T_-}$  denote the jump of the piecewise polynomial  $p \in RT_k(T)$  across an interior edge  $E = \partial T_+ \cap \partial T_-$  shared by the two neighbouring triangles  $T_+$  and  $T_-$ . The Raviart-Thomas finite element space is defined as

$$\begin{aligned} RT_k(\mathcal{T}) &:= \{p \in H(\text{div}, \Omega) \mid p|_T \in RT_k(T) \text{ for all } T \in \mathcal{T}\} \\ &= \left\{ p \in L^2(\Omega) \mid \begin{array}{l} p|_T \in RT_k(T) \text{ for all } T \in \mathcal{T}, \\ [p]_E \cdot \nu_E = 0 \text{ on } E \text{ for all } E \in \mathcal{E}(\Omega) \end{array} \right\}. \end{aligned}$$

For piecewise constant  $\varepsilon > 0$  on  $\mathcal{T}$ , the discrete formulation of  $(\mathbf{D}_\varepsilon)$  reads: Seek  $(u_{\varepsilon h}, \sigma_{\varepsilon h}) \in \mathcal{P}_k(\mathcal{T}) \times RT_k(\mathcal{T})$ , such that for all  $(v_h, \tau_h) \in \mathcal{P}_k(\mathcal{T}) \times RT_k(\mathcal{T})$  it holds that

$$\begin{aligned} (\tau_h, D\Phi_\varepsilon^*(\sigma_{\varepsilon h}))_{L^2(\Omega)} + (u_{\varepsilon h}, \text{div } \tau_h)_{L^2(\Omega)} &= 0, \\ (v_h, \text{div } \sigma_{\varepsilon h})_{L^2(\Omega)} + (f, v_h)_{L^2(\Omega)} &= 0. \end{aligned} \tag{\mathbf{D}_{\varepsilon h}}$$

**4. Existence of exact and discrete solutions.** For discrete form of the original primal problem  $(\mathbf{P})$  on page 5, the discrete primal stress

$$\sigma_h^P := D\Phi_0(\nabla u_h^P) \in \mathcal{P}_0(\mathcal{T}; \mathbb{R}^2)$$

is a piecewise constant solution and existence of  $u_h^P \in \mathcal{P}_1(\mathcal{T}) \cap V$  and the uniqueness of  $\sigma_h^P$  is clarified in [4]. In contrast to the continuous case, the primal and dual solution of the discrete problem do not necessarily coincide and the first step is to prove existence of a discrete solution  $(u_h, \sigma_h)$  of the discrete form of the dual problem  $(\mathbf{D})$ .

Let  $f_h := \Pi_h f \in \mathcal{P}_k(\mathcal{T}) \subseteq L^2(\Omega)$  be the piecewise polynomial  $L^2(\Omega)$  projection of  $f$  with respect to  $\mathcal{T}$  of degree at most  $k \geq 0$  and define

$$\begin{aligned} Q(f, \mathcal{T}) &:= \{\tau_h \in RT_k(\mathcal{T}) \mid f_h + \text{div } \tau_h = 0 \text{ in } \Omega\}, \\ Q(f) &:= \{\tau \in H(\text{div}, \Omega) \mid f + \text{div } \tau = 0\}. \end{aligned}$$

Since  $f \equiv 1 \equiv f_h$  in the optimal design problem (1.1), it holds that  $Q(1, \mathcal{T}) = RT_k(\mathcal{T}) \cap Q(1)$ .

Furthermore, let  $\chi_{Q(f, \mathcal{T})}$  denote the indicator function (cf. [15]) of the convex subset  $Q(f, \mathcal{T}) \subseteq RT_k(\mathcal{T})$ , i.e., for  $\tau_h \in RT_k(\mathcal{T})$ ,

$$\chi_{Q(f, \mathcal{T})}(\tau_h) := \begin{cases} 0 & \text{for } \tau_h \in Q(f, \mathcal{T}), \\ +\infty & \text{otherwise.} \end{cases}$$

THEOREM 4.1 (Existence and uniqueness). *Let  $\varepsilon \geq 0$  piecewise constant with respect to  $\mathcal{T}$ . There exists a unique maximiser  $\sigma_\varepsilon$  of  $E_\varepsilon^*(\tau) := -\int_\Omega \varphi_\varepsilon^*(|\tau|) dx$ , i.e.,*

$$E_\varepsilon^*(\tau) \leq E_\varepsilon^*(\sigma_\varepsilon) \quad \text{for all } \tau \in Q(f).$$

*There exists a unique discrete maximiser  $\sigma_h$  of  $E_0^*$  in  $Q(f, \mathcal{T})$ , i.e.,*

$$E_0^*(\tau_h) \leq E_0^*(\sigma_h) \quad \text{for all } \tau_h \in Q(f, \mathcal{T}), \quad (4.1)$$

*and for all  $\varepsilon \geq 0$  there exists a unique maximiser  $\sigma_{\varepsilon h}$  of  $E_\varepsilon^*$  in  $Q(f, \mathcal{T})$ , i.e.,*

$$-E_\varepsilon^*(\sigma_{\varepsilon h}) \leq -E_\varepsilon^*(\tau_h) + \chi_{Q(f, \mathcal{T})}(\tau_h) \quad \text{for all } \tau_h \in RT_k(\mathcal{T}).$$

*Furthermore, for  $\varepsilon \geq 0$  piecewise constant with respect to  $\mathcal{T}$  there exists some  $u_{\varepsilon h}$ , such that  $(u_{\varepsilon h}, \sigma_{\varepsilon h})$  solves  $(\mathbf{D}_{\varepsilon h})$ . The Lagrange multiplier  $u_{\varepsilon h}$  is unique for  $\varepsilon > 0$ .*

*Proof.* The divergence operator  $\operatorname{div} : H(\operatorname{div}, \Omega) \rightarrow L^2(\Omega)$  is linear and bounded. Hence,  $Q(f)$  is a closed affine subspace. Since,  $\Phi_\varepsilon^*$  is a strongly convex function of quadratic growth on  $H(\operatorname{div}, T)$  for all  $T \in \mathcal{T}$ ,  $-E_\varepsilon^*$  is strongly convex via

$$E_\varepsilon^*(\tau) = -\sum_{T \in \mathcal{T}} \int_T \Phi_\varepsilon^*(\tau) dx$$

in  $H(\operatorname{div}, \Omega)$  and there exists a unique maximiser  $\sigma_\varepsilon$  of  $E_\varepsilon^*$  in  $Q(f)$ . Furthermore, the intersection  $Q(f, \mathcal{T}) := RT_k(\mathcal{T}) \cap Q(f)$  is a closed affine and finite-dimensional subspace and therefore convex and there exists a unique minimiser  $\sigma_h$  of  $-E_0^*$  in  $Q(f, \mathcal{T})$ .

Similar arguments prove that  $\sigma_{\varepsilon h}$  minimises  $-E_\varepsilon^*$  in  $Q(f, \mathcal{T})$ . Hence, [15, Theorem 2.32] verifies

$$0 \in \partial(-E_\varepsilon^*(\sigma_{\varepsilon h})) + \partial\chi_{Q(f, \mathcal{T})}(\sigma_{\varepsilon h}).$$

This proves the existence and uniqueness of a discrete maximiser of  $E_\varepsilon^*$  for all  $\varepsilon \geq 0$ . Furthermore, there exists some  $\xi_h \in \partial(-E_\varepsilon^*(\sigma_{\varepsilon h}))$  with  $-\xi_h \in \partial\chi_{Q(f, \mathcal{T})}(\sigma_{\varepsilon h})$ . The latter reads

$$(-\xi_h, \tau - \sigma_{\varepsilon h})_{L^2(\Omega)} \leq 0 \quad \text{for all } \tau \in Q(f, \mathcal{T}).$$

Since  $2\sigma_{\varepsilon h} - \tau \in Q(f, \mathcal{T})$ , for all  $\tau \in Q(f, \mathcal{T})$ , the reverse inequality

$$(-\xi_h, \sigma_{\varepsilon h} - \tau)_{L^2(\Omega)} \leq 0$$

holds as well. Thus,

$$\xi_h \perp_{L^2(\Omega)} Q(0, \mathcal{T}), \quad \text{i.e., } Q(0, \mathcal{T}) \subseteq \ker \xi_h.$$

It is well-known that the bilinear form  $b : RT_k(\mathcal{T}) \times \mathcal{P}_k(\mathcal{T}) \rightarrow \mathbb{R}$  given by

$$b(q, v) := \int_\Omega v \operatorname{div} q dx \quad \text{for all } q \in RT_k(\mathcal{T}), v \in \mathcal{P}_k(\mathcal{T})$$

fulfils the inf-sup-condition. Therefore the operator  $B$  and its dual  $B^*$ ,

$$\begin{aligned} B : Q(0, \mathcal{T})^\perp &\rightarrow \mathcal{P}_k(\mathcal{T})^*, & q &\mapsto b(q, \cdot); \\ B^* : \mathcal{P}_k(\mathcal{T}) &\rightarrow (Q(0, \mathcal{T})^\perp)^*, & v_h &\mapsto b(\cdot, v_h)|_{Q(0, \mathcal{T})^\perp} \end{aligned}$$

with  $Q(0, \mathcal{T})^\perp \equiv RT_k(\mathcal{T})/Q(0, \mathcal{T})$ , are isomorphisms [1]. For any  $\xi_h \in Q(0, \mathcal{T})^\perp$  there exists a unique Riesz representation  $u_{\varepsilon h} \in \mathcal{P}_0(\mathcal{T})$  with

$$(\xi_h, \tau_h)_{L^2(\Omega)} = B^*(u_{\varepsilon h})(\tau_h) \text{ for all } \tau_h \in Q(f, \mathcal{T})^\perp.$$

This implies that  $(\sigma_{\varepsilon h}, u_{\varepsilon h})$  solves the problem  $(\mathbf{D}_{\varepsilon h})$ . While  $\xi_h$  and thus  $u_{\varepsilon h}$  are unique for  $\varepsilon > 0$ ;  $\xi_h$  and thus  $u_h$  may be non-unique for  $\varepsilon = 0$ .  $\square$

**5. Error Analysis.** This section is devoted to the error analysis of  $(\mathbf{D}_{\varepsilon h})$  by means of the lowest-order Raviart-Thomas finite element space  $RT_k(\mathcal{T})$  for  $k = 0$ .

### 5.1. A priori regularisation error analysis.

**THEOREM 5.1.** *For  $\varepsilon > 0$  piecewise constant with respect to  $\mathcal{T}$  and for  $C_{reg} := \sqrt{C_\mu}/(4\mu_2)$  and  $C_\mu > 0$  from Lemma 2.1, it holds that*

$$\begin{aligned} \|\sigma - \sigma_\varepsilon\|_{L^2(\Omega)} &\leq C_{reg} \|\sqrt{\varepsilon}\sigma_\varepsilon\|_{L^2(\Omega)} \\ &\leq C_{reg} \|\sqrt{\varepsilon}\sigma\|_{L^2(\Omega)} + C_{reg} \|\sqrt{\varepsilon}(\sigma - \sigma_\varepsilon)\|_{L^2(\Omega)}. \end{aligned}$$

For sufficiently small maximal  $\varepsilon_\infty = \|\varepsilon\|_\infty > 0$ , it holds that

$$\|\sigma - \sigma_\varepsilon\|_{L^2(\Omega)} \leq \mathcal{O}(1) \|\sqrt{\varepsilon}\sigma\|_{L^2(\Omega)}.$$

*Proof.* Subsection 2.3.(iv) ensures strong convexity of  $D\Phi_\varepsilon^*$  on all  $T \in \mathcal{T}$ , which implies strong convexity on  $\Omega$ , i.e.,

$$\begin{aligned} 2\mu_2 \|\sigma - \sigma_\varepsilon\|_{L^2(\Omega)}^2 &\leq (D\Phi_\varepsilon^*(\sigma_\varepsilon), \sigma - \sigma_\varepsilon)_{L^2(\Omega)} + \int_\Omega (\Phi_\varepsilon^*(\sigma) - \Phi_\varepsilon^*(\sigma_\varepsilon)) \, dx, \\ 2\mu_2 \|\sigma - \sigma_\varepsilon\|_{L^2(\Omega)}^2 &\leq -(\partial\Phi_0^*(\sigma), \sigma - \sigma_\varepsilon)_{L^2(\Omega)} + \int_\Omega (\Phi_0^*(\sigma_\varepsilon) - \Phi_0^*(\sigma)) \, dx. \end{aligned}$$

Hence, the preceding inequalities hold for all elements of the sets  $D\Phi_\varepsilon^*(\sigma_\varepsilon)$ ,  $\partial\Phi_0^*(\sigma)$  such as  $\nabla u_\varepsilon = D\Phi_\varepsilon^*(\sigma_\varepsilon)$  and  $\nabla u \in \partial\Phi_0^*(\sigma)$ . An integration by parts shows  $(\nabla u, \sigma - \sigma_\varepsilon)_{L^2(\Omega)} = (\nabla u_\varepsilon, \sigma - \sigma_\varepsilon)_{L^2(\Omega)} = 0$ . This implies

$$\begin{aligned} 4\mu_2 \|\sigma - \sigma_\varepsilon\|_{L^2(\Omega)}^2 &\leq \int_\Omega (\Phi_\varepsilon^*(\sigma) - \Phi_0^*(\sigma) + \Phi_0^*(\sigma_\varepsilon) - \Phi_\varepsilon^*(\sigma_\varepsilon)) \, dx \\ &\leq \int_\Omega (\Phi_0^*(\sigma_\varepsilon) - \Phi_\varepsilon^*(\sigma_\varepsilon)) \, dx. \end{aligned}$$

Recall that the Yosida regularisation  $\varphi_\varepsilon^*(t)$  of  $\varphi_0^*(t)$  from Lemma 2.1 with  $C_\mu > 0$  allows for the upper bounds

$$\begin{aligned} 0 &\leq \varphi_0(t)^* - \varphi_\varepsilon^*(t) \leq C_\mu \varepsilon t^2 \text{ for all } t \geq 0, \text{ with } C_\mu > 0, \\ 0 &\leq \int_\Omega (\Phi_0^*(\tau) - \Phi_\varepsilon^*(\tau)) \, dx \leq C_\mu \|\sqrt{\varepsilon}\tau\|_{L^2(\Omega)}^2 \text{ for all } \tau \in L^2(\Omega; \mathbb{R}^2). \end{aligned}$$

Therefore,

$$\begin{aligned} 2\mu_2^{1/2} \|\sigma - \sigma_\varepsilon\|_{L^2(\Omega)} &\leq C_\mu^{1/2} \|\sqrt{\varepsilon}\sigma_\varepsilon\|_{L^2(\Omega)} \\ &\leq C_\mu^{1/2} \|\sqrt{\varepsilon}\sigma\|_{L^2(\Omega)} + C_\mu^{1/2} \|\sqrt{\varepsilon}(\sigma - \sigma_\varepsilon)\|_{L^2(\Omega)}. \end{aligned}$$

Thus, for sufficiently small  $\varepsilon_\infty$ , it holds that

$$\|\sigma - \sigma_\varepsilon\|_{L^2(\Omega)} \leq \frac{1}{2\mu_2^{1/2} C_\mu^{-1/2} - \varepsilon_\infty^{1/2}} \|\sqrt{\varepsilon}\sigma\|_{L^2(\Omega)}. \quad \square$$



**5.2. A priori error analysis of spatial discretisation.** Let  $\Omega_h$  denote the subset of all  $x$  in  $\Omega$  where either  $|\sigma_h(x)| \leq t_1\mu_2 \leq |\sigma(x)|$  or  $|\sigma(x)| \leq t_1\mu_2 \leq |\sigma_h(x)|$ ,

$$\Omega_h := \{x \in \Omega \mid \min\{|\sigma(x)|, |\sigma_h(x)|\} \leq t_1\mu_2 \leq \max\{|\sigma(x)|, |\sigma_h(x)|\}\}.$$

Similarly,  $\Omega_{\varepsilon h}$  denotes the subset of  $\Omega$  of microstructure region for the regularised dual energy density function, i.e.,

$$\Omega_{\varepsilon h} := \left\{ x \in \Omega \mid \begin{array}{l} \exists t \in [t_1\mu_2 + \varepsilon, t_2] \text{ such that} \\ \min\{|\sigma_\varepsilon(x)|, |\sigma_{\varepsilon h}(x)|\} \leq t \leq \max\{|\sigma_\varepsilon(x)|, |\sigma_{\varepsilon h}(x)|\} \end{array} \right\}.$$

The subsequent a priori error estimate leads to an estimate

$$\|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)} \lesssim H^{1/2} \quad (5.1)$$

for the dual solution  $\sigma_\varepsilon \in H^1(\Omega; \mathbb{R}^2)$  and maximal mesh-size  $H := \max_{T \in \mathcal{T}} h_T$ ,  $h_T := |T|^{1/2}$ . For sufficient conditions for the  $H^1$ -regularity of the exact dual solution  $\sigma$  see [6].

**THEOREM 5.2.** *Let  $f \in L^2(\Omega)$  be piecewise constant with respect to  $\mathcal{T}$  and let  $\sigma_\varepsilon \in H^1(\Omega; \mathbb{R}^2)$  the exact dual solution of  $(\mathbf{D}_\varepsilon)$ . Then, the discrete solution  $\sigma_{\varepsilon h}$  of  $(\mathbf{D}_{\varepsilon h})$  on  $\mathcal{T}$  for  $\varepsilon > 0$  satisfies*

$$\|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)} \lesssim H + H^{1/2} |\Omega_{\varepsilon h}|^{1/2}.$$

Before the proof of Theorem 5.2 concludes this section, some remarks are in order.

The numerical investigations in [18] are motivated by the question: Does microstructure arise in this example in the sense that  $\{|\sigma| = t_1\mu_2\}$  has a positive area. This zone of nontrivial Young measure solutions has been observed in the numerical simulations [4, 18] even though its area is usually very small. If the numerical approximation of this area is accurate, then  $|\Omega_{\varepsilon h}| > 0$  and one cannot expect a higher convergence rate than that given in (5.1). Our numerical experiments shall investigate this as well as the preasymptotic behaviour for small  $|\Omega_{\varepsilon h}|/|\Omega| \ll 1$ .

*Proof of Theorem 5.2.* Let  $\mathbf{I}_F : H^1(\Omega; \mathbb{R}^2) \mapsto RT_0(\mathcal{T})$  be Fortin's interpolation operator [2, Section III.3.3] with respect to  $\mathcal{T}$  and defined by

$$\int_E (\sigma_\varepsilon - \mathbf{I}_F \sigma_\varepsilon) \cdot \nu_E \, ds = 0 \text{ for all } E \in \mathcal{E}.$$

Furthermore, let  $\Pi_h : L^2(\Omega) \rightarrow \mathcal{P}_0(\mathcal{T})$  denote the  $L^2$  projection. Besides the commuting diagram property  $\operatorname{div} \mathbf{I}_F \sigma_\varepsilon = \Pi_h \operatorname{div} \sigma_\varepsilon$ , the following estimates [2, Section III.3.3] or [1, Section III.5] hold on  $T \in \mathcal{T}$

$$\|\mathbf{I}_F \sigma_\varepsilon - \sigma_\varepsilon\|_{L^2(T)} \lesssim h_T |\sigma_\varepsilon|_{H^1(T)}. \quad (5.2)$$

The strong monotonicity of  $\Phi_\varepsilon^*$  of Subsection 2.3.(iii) on each  $T \in \mathcal{T}$  yields

$$\mu_2 \|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)}^2 \leq (\mathbf{D}\Phi_\varepsilon^*(\sigma_\varepsilon) - \mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}), \sigma_\varepsilon - \sigma_{\varepsilon h})_{L^2(\Omega)}. \quad (5.3)$$

Since  $\sigma_{\varepsilon h}, \mathbf{I}_F \sigma_\varepsilon \in Q(f) = Q(f, \mathcal{T})$  and  $\mathbf{D}\Phi_\varepsilon^*(\sigma_\varepsilon) = \nabla u_\varepsilon$  the  $L^2$ -orthogonalities  $\mathbf{D}\Phi_\varepsilon^*(\sigma_\varepsilon) \perp_{L^2(\Omega)} \mathbf{I}_F(\sigma_\varepsilon - \sigma_{\varepsilon h})$  and  $u_{\varepsilon h} \perp_{L^2(\Omega)} \operatorname{div}(\sigma_{\varepsilon h} - \mathbf{I}_F \sigma_\varepsilon)$  hold. Hence, (5.2)-(5.3) prove

$$\mu_2 \|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)}^2 \leq (\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}), \mathbf{I}_F \sigma_\varepsilon - \sigma_\varepsilon)_{L^2(\Omega)} - (u_{\varepsilon h}, \operatorname{div}(\sigma_\varepsilon - \mathbf{I}_F \sigma_\varepsilon))_{L^2(\Omega)}.$$

Furthermore, since  $\operatorname{div}(\mathbf{I}_F \sigma_\varepsilon - \sigma_\varepsilon) = f - f_h = 0$  for piecewise constant  $f$  and  $0 = (u_{\varepsilon h}, \operatorname{div}(\sigma_\varepsilon - \mathbf{I}_F \sigma_\varepsilon))_{L^2(\Omega)}$  the following estimate holds

$$\begin{aligned} \mu_2 \|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)}^2 &\leq (\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}), \mathbf{I}_F \sigma_\varepsilon - \sigma_\varepsilon)_{L^2(\Omega)} + (u_h, \operatorname{div}(\mathbf{I}_F \sigma_\varepsilon - \sigma_\varepsilon))_{L^2(\Omega)} \\ &= (\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}) - \mathbf{D}\Phi_\varepsilon^*(\sigma_\varepsilon), \mathbf{I}_F \sigma_\varepsilon - \sigma_\varepsilon)_{L^2(\Omega)}. \end{aligned}$$

Cauchy-Schwarz' inequality, Subsection 2.3.(vi) and the estimates of Fortin's interpolation with  $C_F > 0$  lead to

$$\begin{aligned} \mu_2 \|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)}^2 &\leq \|\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}) - \mathbf{D}\Phi_\varepsilon^*(\sigma_\varepsilon)\|_{L^2(\Omega)} \|\mathbf{I}_F \sigma_\varepsilon - \sigma_\varepsilon\|_{L^2(\Omega)} \\ &\leq \left( \|\delta_\varepsilon(\sigma_\varepsilon, \sigma_{\varepsilon h})\|_{L^2(\Omega)} + 1/\mu_1 \|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)} \right) C_F H |\sigma_\varepsilon|_{H^1(\Omega)}. \end{aligned}$$

With  $\Omega_{\varepsilon h}$  from the beginning of this subsection and Subsection 2.3.(vi) in the sense of

$$\|\delta_\varepsilon(\sigma_\varepsilon, \sigma_{\varepsilon h})\|_{L^2(\Omega)} \leq |\Omega_{\varepsilon h}| (t_2 - t_1) \lesssim 1,$$

one concludes

$$\begin{aligned} \mu_2 \|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)}^2 &\leq H C_F (t_2 - t_1) |\Omega_{\varepsilon h}| |\sigma_\varepsilon|_{H^1(\Omega)} \\ &\quad + H C_F / \mu_1 \|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)} |\sigma_\varepsilon|_{H^1(\Omega)}. \end{aligned}$$

Young's inequality proves the assertion

$$\|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)}^2 \lesssim H |\Omega_{\varepsilon h}| |\sigma_\varepsilon|_{H^1(\Omega)} + H^2 |\sigma_\varepsilon|_{H^1(\Omega)}^2 \lesssim H |\Omega_{\varepsilon h}| + H^2. \quad \square$$

### 5.3. A posteriori error analysis.

**THEOREM 5.3.** *For the exact and discrete solutions  $\sigma_\varepsilon \in H^1(\Omega; \mathbb{R}^2)$  and  $\sigma_{\varepsilon h} \in RT_0(\mathcal{T})$  of  $(\mathbf{D}_\varepsilon)$  and  $(\mathbf{D}_{\varepsilon h})$ ,  $\varepsilon > 0$  with piecewise constant right-hand side  $f \in L^2(\Omega)$  with respect to  $\mathcal{T}$  and for  $C := C_C(1 + C_{\text{reg}}\sqrt{\varepsilon_\infty})$  and for positive constants  $C_{\text{reg}}$  of Theorem 5.1 and  $C_C$  of Clément's interpolation, it holds*

$$\begin{aligned} \|\sigma - \sigma_{\varepsilon h}\|_{L^2(\Omega)} &\leq C_{\text{reg}} \|\sqrt{\varepsilon} \sigma_{\varepsilon h}\|_{L^2(\Omega)} + 1/\mu_2 \min_{v \in V} \|\mathbf{D}\Phi_\varepsilon^*(|\sigma_{\varepsilon h}|) - \nabla v\|_{L^2(\Omega)} \end{aligned} \quad (5.4)$$

$$\begin{aligned} &\leq C_{\text{reg}} \|\sqrt{\varepsilon} \sigma_{\varepsilon h}\|_{L^2(\Omega)} + C/\mu_2 \left( \sum_{E \in \mathcal{E}} \left\| h_E^{1/2} [\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h})] \cdot \tau_E \right\|_{L^2(E)} \right. \\ &\quad \left. + \sum_{T \in \mathcal{T}} \|h_T \operatorname{curl} \mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h})\|_{L^2(T)} \right). \end{aligned} \quad (5.5)$$

*Proof.* The triangle inequality and the estimates of Theorem 5.1 reveal

$$\begin{aligned} \|\sigma - \sigma_{\varepsilon h}\|_{L^2(\Omega)} &\leq \|\sigma - \sigma_\varepsilon\|_{L^2(\Omega)} + \|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)} \\ &\leq C_{\text{reg}} \|\sqrt{\varepsilon} \sigma_\varepsilon\|_{L^2(\Omega)} + \|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)} \\ &\leq C_{\text{reg}} \|\sqrt{\varepsilon} \sigma_{\varepsilon h}\|_{L^2(\Omega)} + \|(1 + C_{\text{reg}}\sqrt{\varepsilon})(\sigma_\varepsilon - \sigma_{\varepsilon h})\|_{L^2(\Omega)}. \end{aligned}$$

Furthermore, the inequality of Subsection 2.3.(iii) leads for  $\varepsilon > 0$  to

$$\mu_2 \|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)}^2 \leq (\mathbf{D}\Phi_\varepsilon^*(\sigma_\varepsilon) - \mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}), \sigma_\varepsilon - \sigma_{\varepsilon h})_{L^2(\Omega)}.$$

Since  $\nabla u_\varepsilon = \mathbf{D}\Phi_\varepsilon^*(\sigma_\varepsilon)$ , the right-hand side equals

$$(\nabla u_\varepsilon - \mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}), \sigma_\varepsilon - \sigma_{\varepsilon h})_{L^2(\Omega)}.$$

An integration by parts with  $u_\varepsilon \in V$  shows that this equals

$$(u_\varepsilon, \operatorname{div}(\sigma_{\varepsilon h} - \sigma_\varepsilon))_{L^2(\Omega)} - (\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}), \sigma_\varepsilon - \sigma_{\varepsilon h})_{L^2(\Omega)}.$$

Since  $\operatorname{div}(\sigma_\varepsilon - \sigma_{\varepsilon h}) = f - f_h \equiv 0$  for piecewise constant  $f$ , the first term vanishes. The same argument for any  $v \in V$  results in

$$\begin{aligned} \mu_2 \|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)}^2 &\leq (\nabla v - \mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}), \sigma_\varepsilon - \sigma_{\varepsilon h})_{L^2(\Omega)} \\ &\leq \|\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}) - \nabla v\|_{L^2(\Omega)} \|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)}. \end{aligned}$$

Hence,

$$\mu_2 \|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)} \leq \min_{v \in V} \|\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}) - \nabla v\|_{L^2(\Omega)}.$$

Define  $\tilde{v} := \operatorname{argmin}_{v \in V} \|\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}) - \nabla v\|_{L^2(\Omega)}$  so that  $\tilde{v} \in V$  satisfies

$$(\nabla \tilde{v}, \nabla w)_{L^2(\Omega)} = (\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}), \nabla w)_{L^2(\Omega)} \quad \text{for all } w \in V. \quad (5.6)$$

The Helmholtz decomposition [12] of  $\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h})$  in  $\alpha \in V$  and  $\beta \in H^1(\Omega)/\mathbb{R}$  reads

$$\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}) = \nabla \alpha + \operatorname{Curl} \beta \quad (5.7)$$

with an orthogonal split  $(\nabla \alpha, \operatorname{Curl} \beta)_{L^2(\Omega)} = 0$ . Hence,

$$\mu_2 \|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)} \leq \|\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}) - \nabla \tilde{v}\|_{L^2(\Omega)} = \|\operatorname{Curl} \beta\|_{L^2(\Omega)}.$$

For  $z \in \mathcal{N}$  define  $\omega_z := \{T \in \mathcal{T} \mid z \in T\}$  as the patch to the node  $z$ . Define the nodal function  $\phi_z \in \mathcal{S}_1(\mathcal{T})$  by  $\phi_z(z) := 1$  for  $z \in \mathcal{N}$  and  $\phi_z(y) = 0$  for  $y \in \mathcal{N} \setminus \{z\}$ .

Let  $J : H^1(\Omega) \rightarrow \mathcal{S}_1(\mathcal{T})$  the Clément-interpolation operator and  $J\beta$  the interpolator of  $\beta$  given by [2]

$$J(\beta) := \sum_{z \in \mathcal{N}} \beta_z \phi_z \quad \text{with } \beta_z := \begin{cases} |\omega_z|^{-1} \int_{\omega_z} \beta \, dx & \text{for all } z \in \mathcal{N} \setminus \partial\Omega, \\ 0 & \text{for all } z \in \mathcal{N} \cap \partial\Omega. \end{cases}$$

Thus,  $\operatorname{Curl} J\beta \in \mathcal{P}_0(\mathcal{T}) \cap H(\operatorname{div}, \Omega) \subset RT_0(\mathcal{T})$  and for  $\beta \in H^1(\Omega)$  the following estimates hold [10]

$$\|\nabla J(\beta)\|_{L^2(\Omega)} + \|h_{\mathcal{T}}^{-1}(\beta - J(\beta))\|_{L^2(\Omega)} + \|h_{\mathcal{E}}^{-1/2}(\beta - J(\beta))\|_{L^2(\cup \mathcal{E})} \lesssim \|\nabla \beta\|_{L^2(\Omega)}.$$

Let  $[J\beta] \cdot \nu$  denote the jump of  $J\beta$  in normal direction across (and  $[J\beta] \cdot \tau$  in tangential direction along) the edges in  $\mathcal{E}$ . Hence,  $[[\operatorname{Curl} J\beta] \cdot \nu]_{\mathcal{E}} = [[J\beta] \cdot \tau]_{\mathcal{E}} = 0$ .

Since  $\text{Curl } J\beta \perp \nabla H_0^1(\Omega)$  and  $(\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}), q_h) = (-u_{\varepsilon h}, \text{div } q_h)$  hold for all  $q_h \in RT_0(\mathcal{T})$ , the orthogonal split  $\text{Curl } \beta \perp \text{Curl } J\beta$  is verified

$$\begin{aligned} (\text{Curl } \beta, \text{Curl } J\beta)_{L^2(\Omega)} &= (\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}) - \nabla \alpha, \text{Curl } J\beta)_{L^2(\Omega)} \\ &= -(u_{\varepsilon h}, \text{div } \text{Curl } J\beta)_{L^2(\Omega)} = 0. \end{aligned}$$

Let  $C_C > 0$  be a constant from Clément's interpolation error estimates. The orthogonality  $\text{Curl } \beta \perp \text{Curl } J\beta$  yields

$$\begin{aligned} \|\text{Curl } \beta\|_{L^2(\Omega)}^2 &= (\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}) - \nabla \alpha, \text{Curl } (\beta - J\beta))_{L^2(\Omega)} \\ &= \sum_{T \in \mathcal{T}} \left( -(\text{curl } \mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}), \beta - J\beta)_{L^2(T)} \right. \\ &\quad \left. + (\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}) \cdot \tau, \beta - J\beta)_{L^2(\partial T)} \right) \\ &\leq \sum_{T \in \mathcal{T}} \|h_T \text{curl } \mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h})\|_{L^2(T)} \|h_T^{-1} (\beta - J\beta)\|_{L^2(T)} \\ &\quad + \sum_{E \in \mathcal{E}} \left\| h_E^{1/2} [\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h})] \cdot \tau_E \right\|_{L^2(E)} \left\| h_E^{-1/2} (\beta - J\beta) \right\|_{L^2(E)} \\ &\leq C_C \left( \sum_{T \in \mathcal{T}} \|h_T \text{curl } \mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h})\|_{L^2(T)}^2 \right. \\ &\quad \left. + \sum_{E \in \mathcal{E}} \left\| h_E^{1/2} [\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h})] \cdot \tau_E \right\|_{L^2(E)}^2 \right)^{1/2} \|\nabla \beta\|_{L^2(\Omega)}. \end{aligned}$$

Finally,  $\|\nabla \beta\|_{L^2(\Omega)} = \|\text{Curl } \beta\|_{L^2(\Omega)}$  shows the assertion.  $\square$

**REMARK 5.4.** *Given the definition of  $\varphi_0^*$  from the variational formulation of the optimal design example in Section 2.1 and its regularisation  $\varphi_\varepsilon^*$ , one observes that  $\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h})$  is a Raviart-Thomas element shape function in the interior of each material  $\Omega_1$  and  $\Omega_2$ , where  $\text{curl } \mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}) = 0$ .*

*Hence, the elements  $T$  in a neighbourhood of the contact zone of the two materials exclusively contribute to  $\sum_{T \in \mathcal{T}} \|h_T \text{curl } \mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h})\|_{L^2(T)}^2$  and the jump term in (5.5) may dominate the error estimator.*

**REMARK 5.5.** *The right-hand side in (5.4)-(5.5) is expected to be sharp in the sense that the arguments are known to lead to efficient error control in many applications of mixed FEM. Standard techniques for an efficiency proof, however, encounter the non-smoothness of  $\mathbf{D}\Phi_\varepsilon^*$  as  $\varepsilon \searrow 0$ .*

**6. Numerical Experiments.** This section is devoted to numerical experiments for the degenerate variational problem in its dual discrete mixed formulation  $(\mathbf{D}_{\varepsilon h})$  based on Raviart-Thomas FEM in comparison to the discrete solutions of  $(\mathbf{P})$  in [4] with  $P_1$ -FEM on the domains of Figure 6.1 the square, the L-shaped domain and the octagon. In all of these examples the loads and the boundary conditions are given by  $f \equiv 1$  and  $u_D \equiv 0$ . The material distribution is set to  $\xi = 0.5$ , thus both materials fill half of the domain, with the material parameters  $\mu_1 = 1 < \mu_2 = 2$ .

**6.1. Preliminary Remarks.** In the variational formulation of the primal problem the Lagrange-multiplier for the material distribution is  $\lambda$ , cf. [4] for a motivation and the computation of the optimal values shown in Figure 6.1.

However, an approximation of the exact primal and conjugated energy seems very discerning. The arduousness lies in the fact, that extrapolation is significant

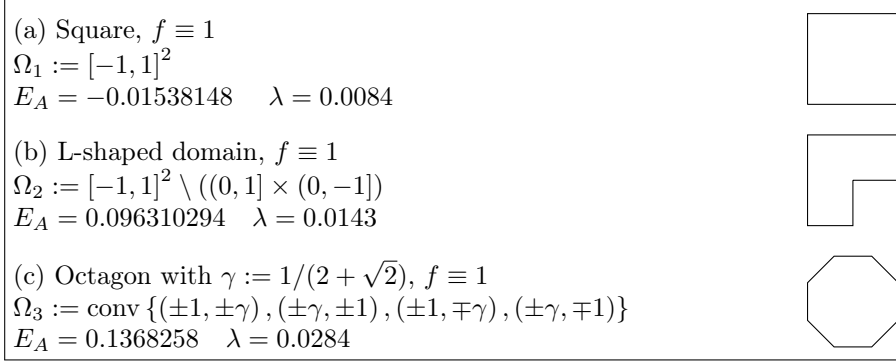


FIG. 6.1. Domains of the four numerical benchmarks, its extrapolated energies  $E_A$  (rounded off), and optimal  $\lambda$

only on uniform meshes, while for uniformly refined meshes the contact zone of both materials in the cross section is not adequately resolved. Thus, only a low number of digits appears trustworthy. This leads to objectionable effects in the convergence graphs of the approximation of the energy error. The extrapolation of sequences of the dual energy

$$E_\varepsilon^* := \int_{\Omega} \varphi_\varepsilon^*(|\sigma_h|), \text{ for } \varepsilon \geq 0$$

on uniform meshes based on the dual mixed formulation ( $\mathbf{D}_{\varepsilon h}$ ) appears non-reliable. The listed extrapolated energies  $E_A$  and  $E_A^*$  have been calculated by some Aitken extrapolation algorithm on uniform refined meshes, generated by  $\mathcal{S}_1$  conforming FEM based on the discrete form of the primal problem ( $\mathbf{P}$ ) as in [4].

The analysis of Section 5 motivates the following two estimators  $\eta_H$  and  $\eta_R$ , for  $\varepsilon > 0$ ,

$$\eta_H^2 := \min_{v \in \mathcal{S}_{1,0}(\mathcal{T})} \|\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h}) - \nabla v\|_{L^2(\mathcal{T})}^2, \quad (6.1)$$

$$\eta_E^2 := \sum_{E \in \mathcal{E}} \left\| h_E^{1/2} [\mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h})] \cdot \tau_E \right\|_{L^2(E)}^2,$$

$$\eta_R^2 := \eta_E^2 + \sum_{T \in \mathcal{T}} \|h_T \text{curl } \mathbf{D}\Phi_\varepsilon^*(\sigma_{\varepsilon h})\|_{L^2(T)}^2. \quad (6.2)$$

Since the exact solution is not available, the convergence behaviour of the estimators (6.1) and (6.2) are compared for adaptive and uniform mesh refinement.

The algorithm presented in the sequel solves ( $\mathbf{D}_{\varepsilon h}$ ) and decreases  $\varepsilon$  locally for an accurate computation.

**ALGORITHM 6.1.** *Input:* shape regular triangulation  $\mathcal{T}_0$ , initial value  $(u_{\varepsilon 0}, \sigma_{\varepsilon 0})$ , regularisation parameters  $\alpha$  and  $\beta$ , tolerance  $0 < \text{Tol}$ . Set  $\eta_0 := \infty$ ,  $\ell := 1$ . **WHILE**  $\eta_{\ell-1} \geq \text{Tol}$  **DO** (i)-(v):

(i) Create new triangulation  $\mathcal{T}_\ell$  corresponding to the estimated error  $\eta_\ell$ .

(ii) Prolongate  $(u_{\varepsilon \ell-1}, \sigma_{\varepsilon \ell-1})$  to  $\mathcal{T}_\ell$  to get an initial value  $(u_{\varepsilon \ell}^0, \sigma_{\varepsilon \ell}^0)$ .

(iii) Update regularisation parameter  $\varepsilon|_T = \alpha h_T^\beta$ ,  $T \in \mathcal{T}_\ell$ .

(iv) Compute solution  $(u_{\varepsilon \ell}, \sigma_{\varepsilon \ell})$  of ( $\mathbf{D}_{\varepsilon h}$ ) with Gauss-Newton method provided in Matlab's `fsolve` and initial value  $(u_{\varepsilon \ell}^0, \sigma_{\varepsilon \ell}^0)$ .

(v) Calculate estimated error  $\eta_\ell$  of the solution  $\sigma_{\varepsilon\ell}$  and set  $\ell = \ell + 1$ ;  
 Output: Approximation of the solution of  $(\mathbf{D}_{\varepsilon h})$ .

REMARK 6.2. The estimates  $\|\sigma - \sigma_\varepsilon\|_{L^2(\Omega)} \lesssim \|\sqrt{\varepsilon}\sigma_\varepsilon\|_{L^2(\Omega)}$  and  $\|\sigma_\varepsilon - \sigma_{\varepsilon h}\|_{L^2(\Omega)} \lesssim H^{1/2}$  from Theorems 5.1 and 5.2 suggest  $\varepsilon = \mathcal{O}(h)$ . To find appropriate values of  $\alpha$  and  $\beta$  in the ansatz  $\varepsilon|_T := \alpha h_T^\beta$ , Algorithm 6.1 has run for various choices of  $\alpha$  and  $\beta$  on the test setting on the unit square  $\Omega$  with the exact solution

$$\tilde{u}(x, y) := xy(1-x)(1-y). \quad (6.3)$$

The error estimators  $\eta_H$  and  $\eta_R$ , the exact stress errors  $\|\sigma - \sigma_{\varepsilon h}\|_{L^2(\Omega)}$ , and the square root of the energy error  $\delta_\ell := |E(\sigma_{\varepsilon h}) - E_A|$  have been evaluated for various parameters and let to the conjecture that  $\alpha = \beta = 1$  is a proper choice for the regularisation parameter  $\varepsilon = h_T$  in Algorithm 6.1.

**6.2. Optimal Design on different domains.** To analyse the quality of results produced by Algorithm 6.1, it is applied to the examples introduced in Figure 6.1. For each domain the exact energy is approximated by an Aitken extrapolation of the discrete energy, this extrapolated energy  $E_A$  is given in Figure 6.1.

For each domain, the subsequent Figures show the approximated optimal volume fraction

$$\begin{cases} 0 & \text{for } 0 \leq |\nabla u_{\varepsilon h}| \leq t_1, \\ (|\nabla u_{\varepsilon h}| - t_1)/(t_2 - t_1) & \text{for } t_1 \leq |\nabla u_{\varepsilon h}| \leq t_2, \\ 1 & \text{for } t_2 \leq |\nabla u_{\varepsilon h}| \end{cases}$$

of each material indicated by regions colored *red* and *blue* on the left-hand side and the region of microstructure where both materials are present in *black* on the right-hand side. The estimated errors  $\eta_H$ ,  $\eta_R$  and square root of the energy error for sequences of uniform and adaptively generated triangulations are plotted in dependence of the number of degrees of freedom. Furthermore a subsequence of the  $\eta_H$ -adaptively generated grids is shown. For the squared domain those results are presented in Figures 6.2-6.4.

The error estimators and square root of extrapolated energy errors  $\delta_\ell^{1/2} := |E(\sigma_{\varepsilon h}) - E_A|^{1/2}$  and  $\delta_\ell^{*1/2} := |E^*(\sigma_{\varepsilon h}) - E_A^*|^{1/2}$  are plotted in a double logarithmic scaling in dependence of the number of degrees of freedom (ndof).

With the L-shaped domain and adaptive refinement the rate of convergence compared to uniform refinement is improved significantly. Apparently, the area  $\{x \in \Omega \mid t_1 < |\nabla u_{\varepsilon h}(x)| < t_2\}$ , where both materials are present seems to be very small.

If the parameter  $\xi$ , which influences the volume fraction of each material, is chosen in a way such that the contact zone is just a boundary, i.e., there is no subdomain where both materials are present, the error estimators show optimal convergence, cf. Figure 6.15 for the L-shaped domain and  $\xi = 0.8$ .

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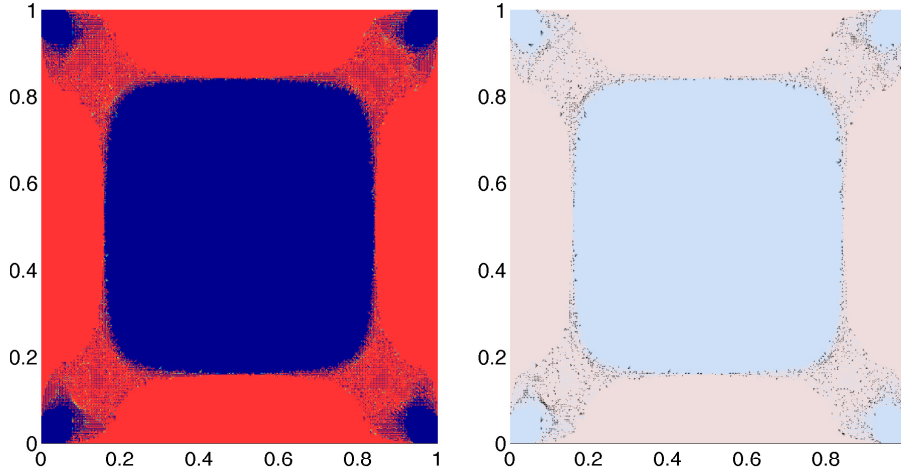


FIG. 6.2. (LHS) Volume fraction for the two materials (blue and red) for adaptive refinement for the Square generated by Algorithm 6.1 and  $\eta_R$ . (RHS) The region of microstructure as approximated mixed zone where both materials are present (black).

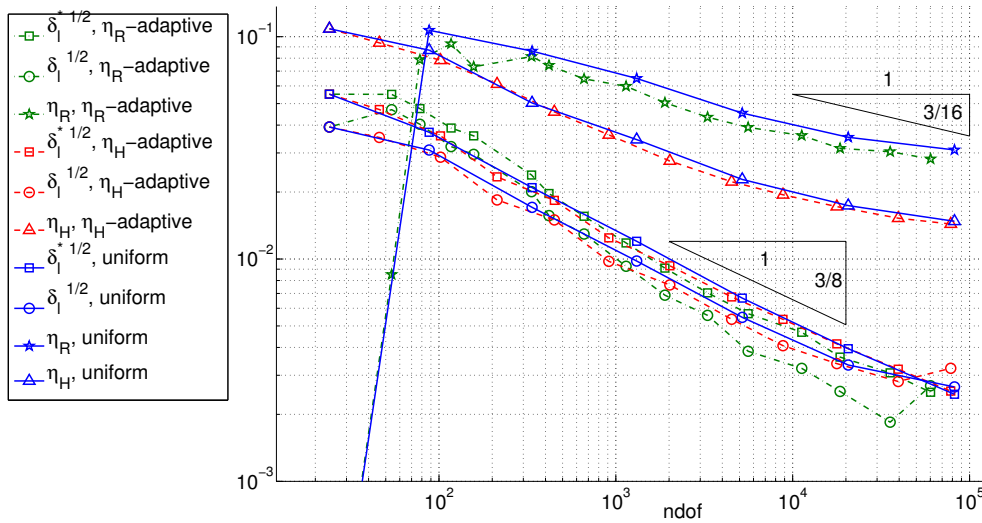


FIG. 6.3. Convergence history; error estimators and extrapolated energy error for the square.

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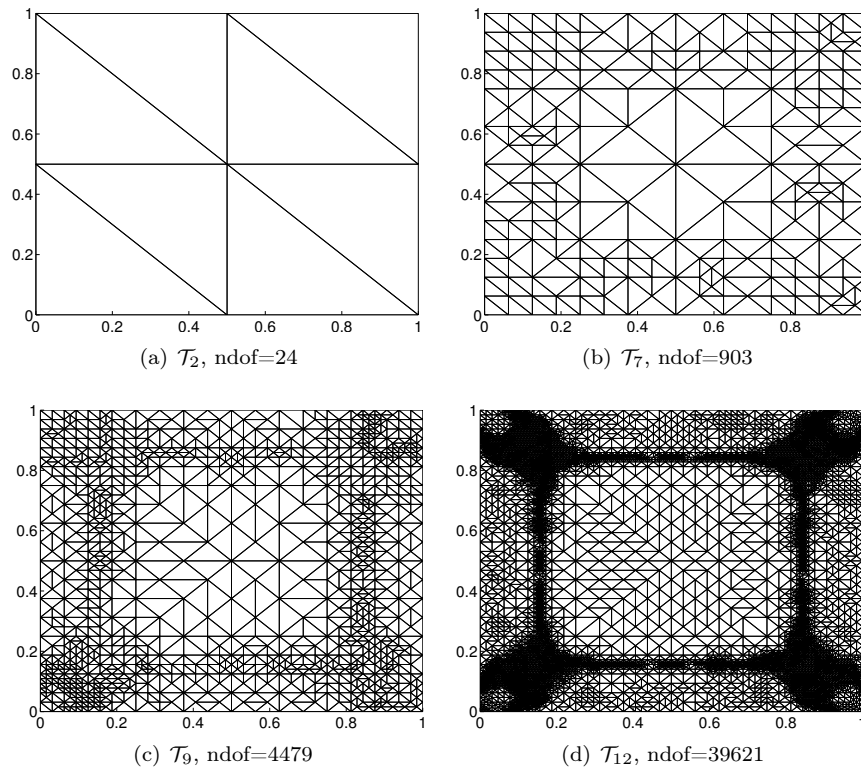


FIG. 6.4. Sequence of meshes, generated by Algorithm 6.1 and  $\eta_H$  for the square.

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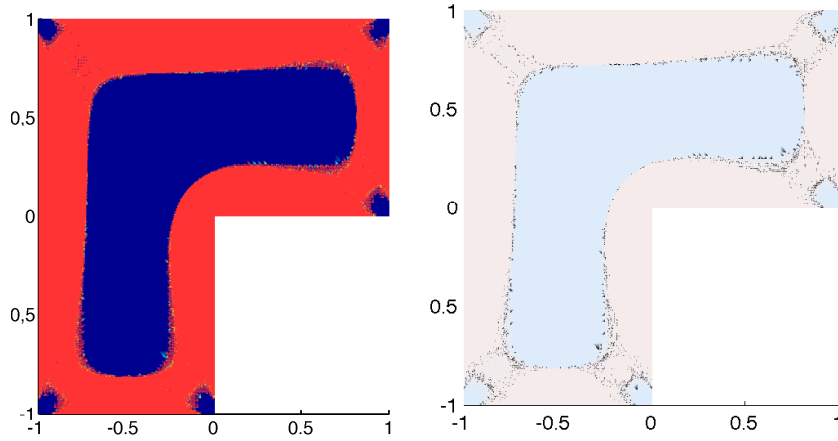


FIG. 6.5. Volume fraction for uniform refinement for the L-shaped domain generated by Algorithm 6.1 and  $\eta_R$  (LHS) and its microstructure (RHS).

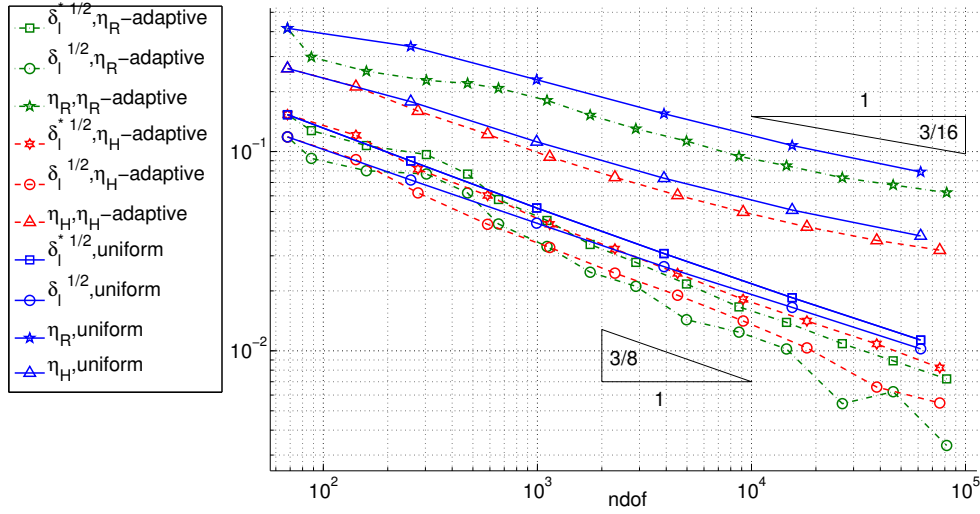


FIG. 6.6. Convergence history; error estimators and extrapolated energy error for the L-shaped domain.

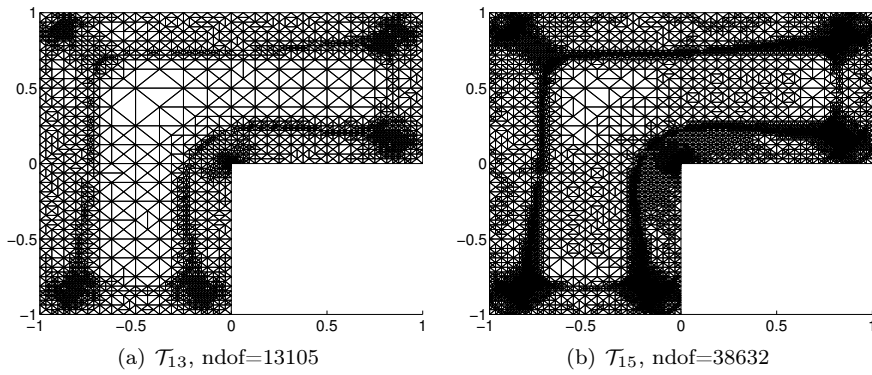


FIG. 6.7. Subsequence of meshes of the L-shaped domain generated by Algorithm 6.1 and  $\eta_R$ .

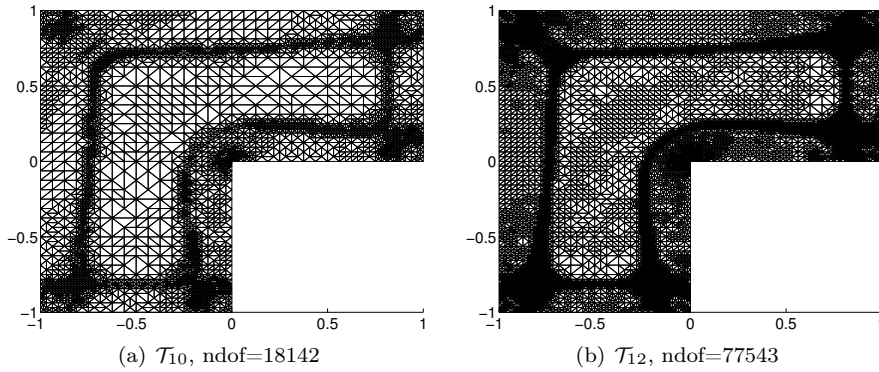


FIG. 6.8. Subsequence of meshes of the L-shaped domain generated by Algorithm 6.1 and  $\eta_H$ .

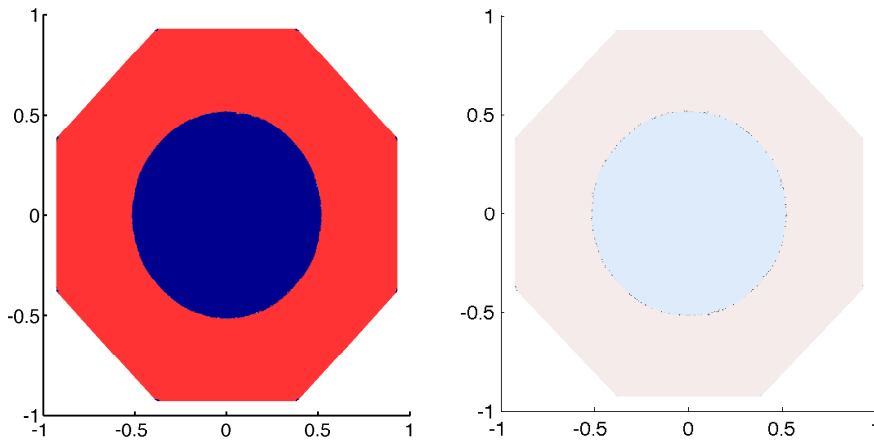


FIG. 6.9. Volume fraction for uniform refinement for the octagon generated by Algorithm 6.1 and  $\eta_R$ .

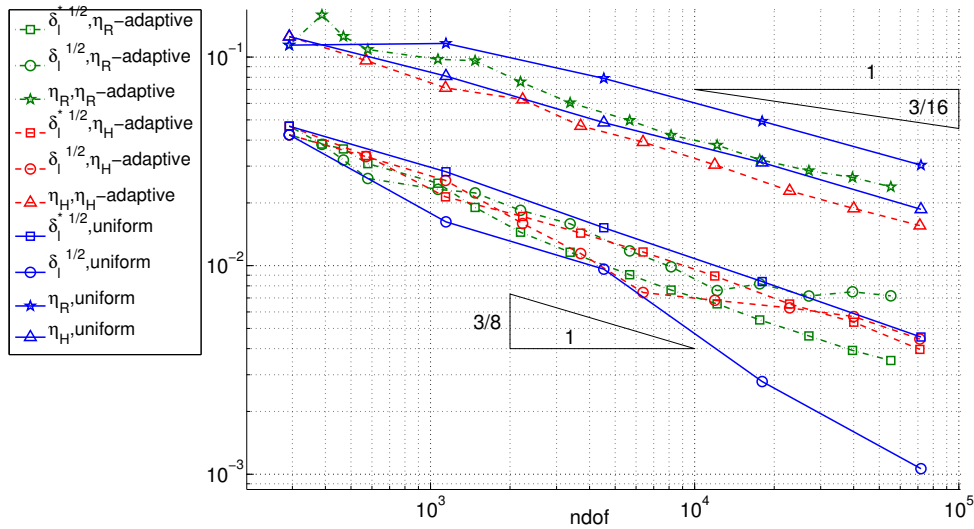


FIG. 6.10. Convergence history; error estimators and extrapolated energy error for the octagon.

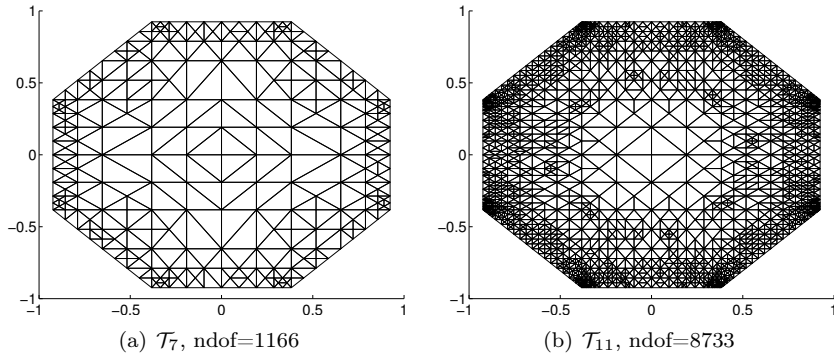


FIG. 6.11. Subsequence of meshes, generated by Algorithm 6.1 and  $\eta_R$  for the octagon.

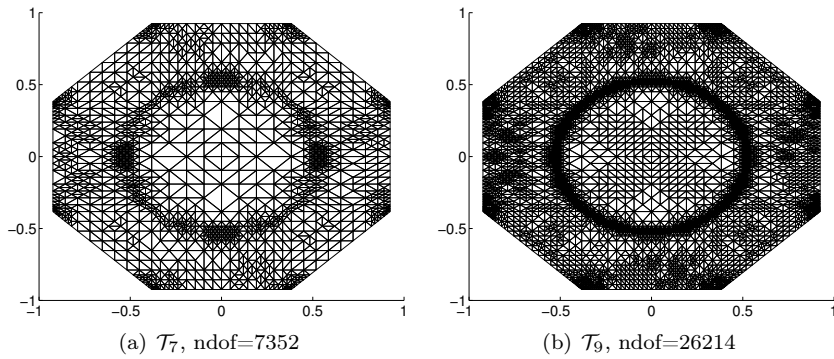


FIG. 6.12. Subsequence of meshes, generated by Algorithm 6.1 and  $\eta_H$  for the octagon.

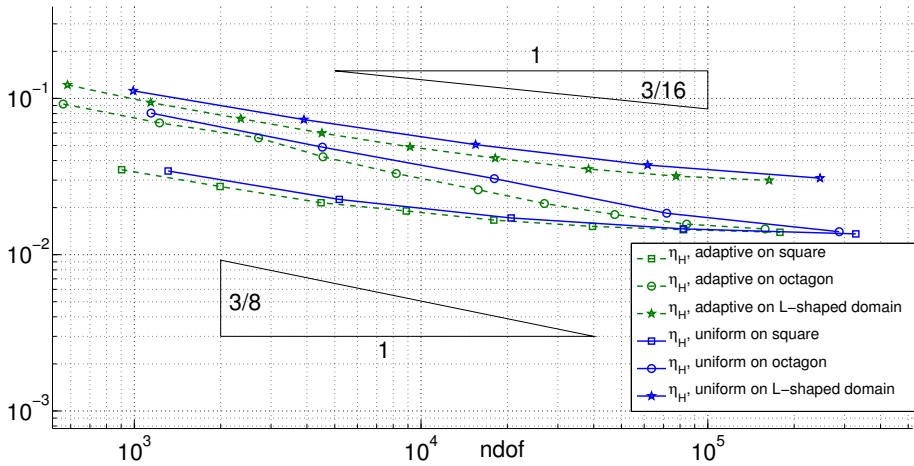


FIG. 6.13. Convergence history of  $\eta_H$  for adaptive and uniform refinement on all benchmark domains.

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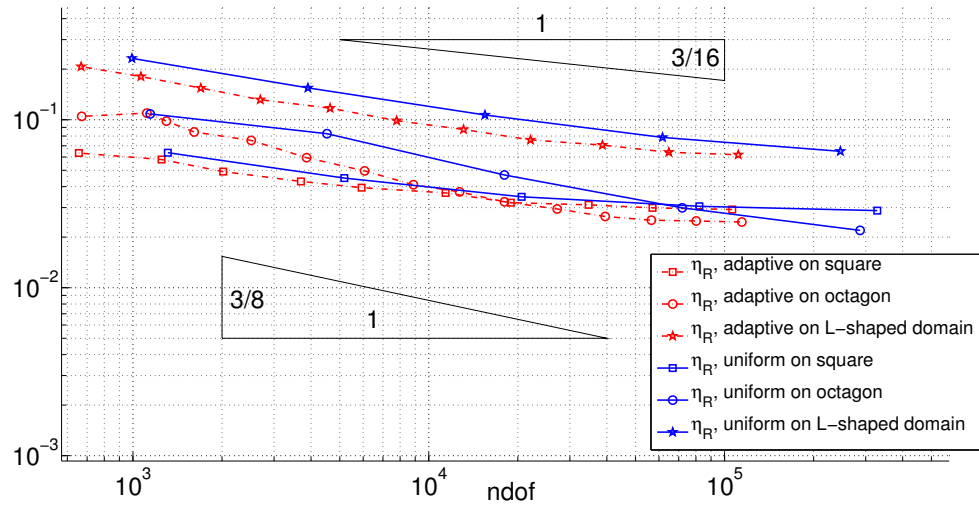
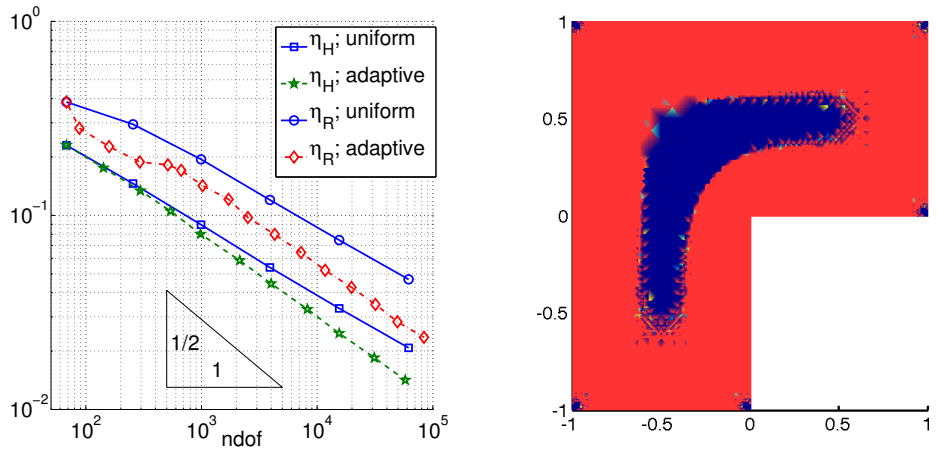


FIG. 6.14. Convergence history  $\eta_R$  for adaptive and uniform refinement on all benchmark domains.



(a) Convergence history for uniform and adaptive refinement for the error estimators. (b) Volume fraction for the L-shaped domain.

FIG. 6.15. L-shaped domain for a material distribution of  $\xi = 0.8$ .